

# Introduction & Administrivia

EECE695D: Efficient ML Systems

Spring 2025

# What happened?

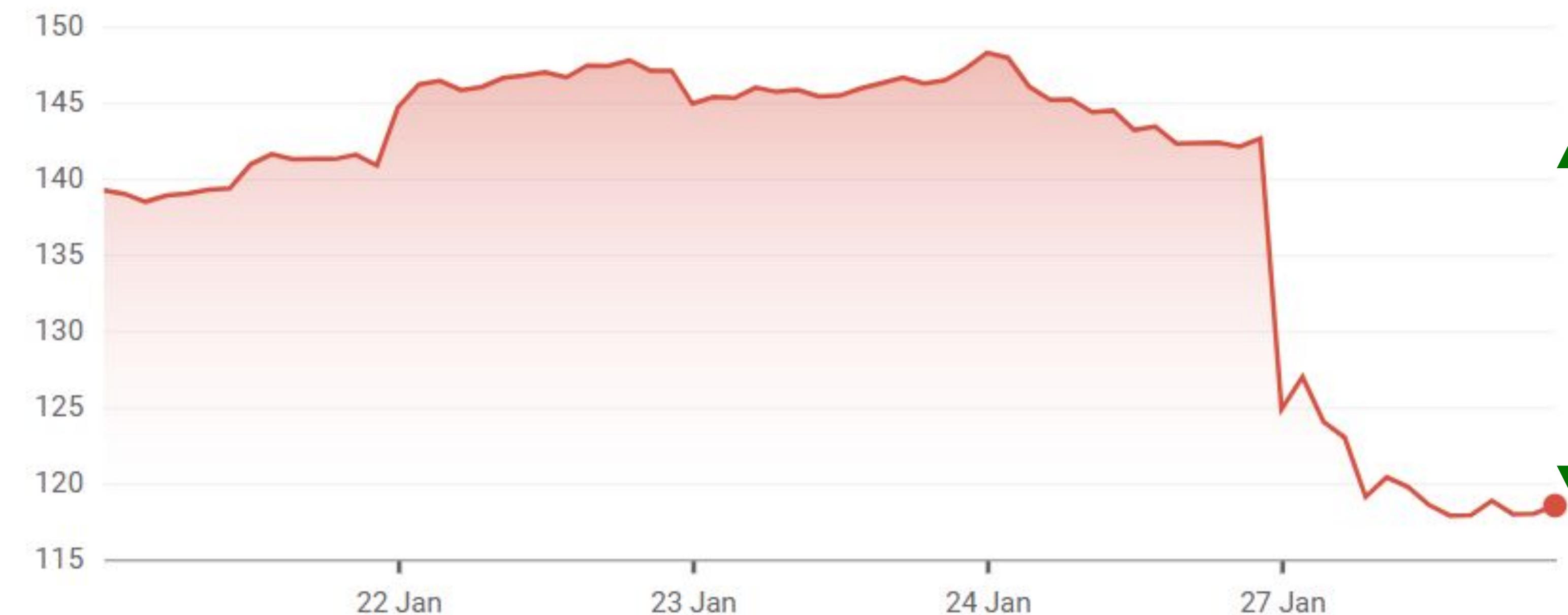
NVIDIA Corp

\$118.58 ↓14.83% -20.64 5 D

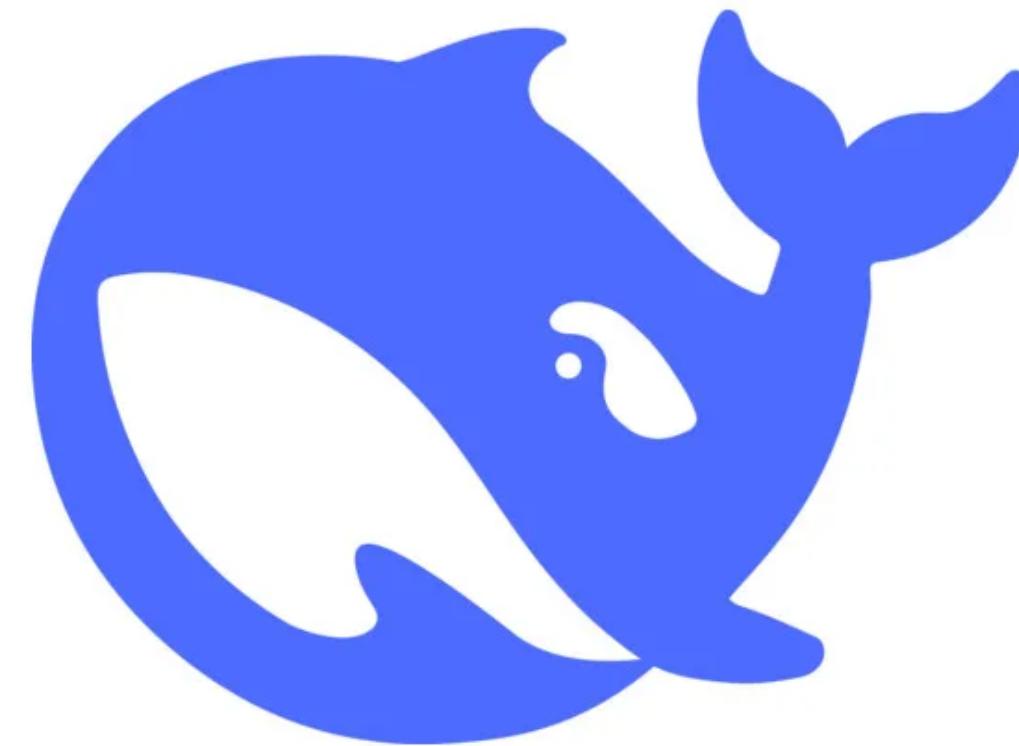
After hours: \$121.27 (↑2.27%) +2.69

Closed: 27 Jan, 19:30:04 UTC-5 · USD · NASDAQ · Disclaimer

1 D    5 D    1 M    6 M    YTD    1Y    5Y    MAX



\$560B = 2.7 \* R&D budget of  
Korean government



## Semiconductor



-17%



-13%



-10%



-7%



-7%

## IT Hardware



-22%



-9%

## Data Centers



-9%



-4%

## Electricity



-28%



-22%



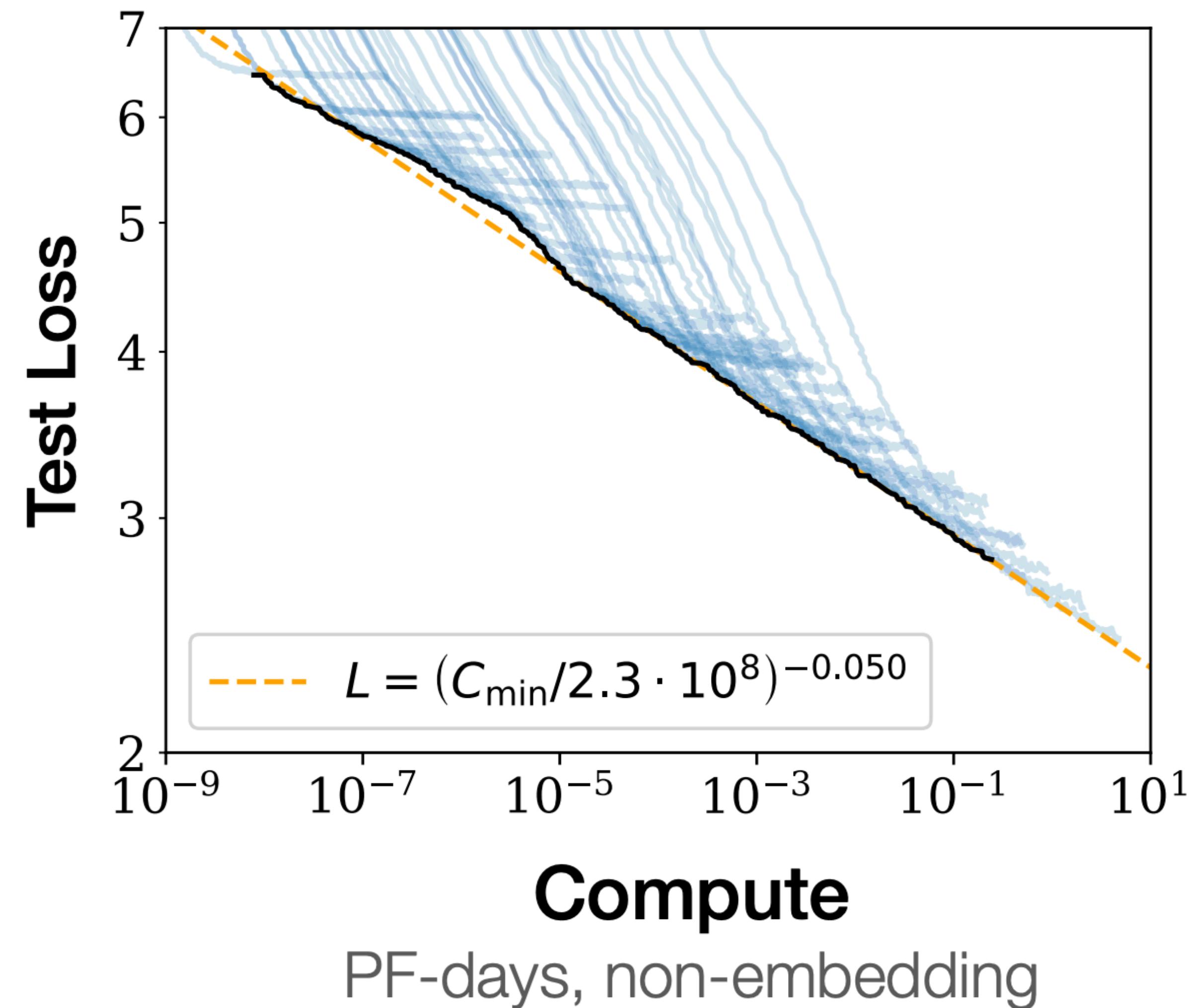
-21%

# Why was DeepSeek so disruptive?

- New Capability? No!
- Cheaper Cost? Yes ;)
  - \$6M for LLM pre-training  
(\$100M for GPT-4, reportedly)
  - 2,000 H800 GPUs  
(6,000–10,000 H100 for GPT-o1, reportedly)

# That means...

- Given the same budget, we can
  - use larger models
  - train longer
  - use more data
  - ... and get **better performance!**



# Also, the inference cost

- On the **inference** side, cheaper inference cost means:

- Cheaper LLM subscription
- Smarter on-device AI
- Better reasoning
- Better generation quality



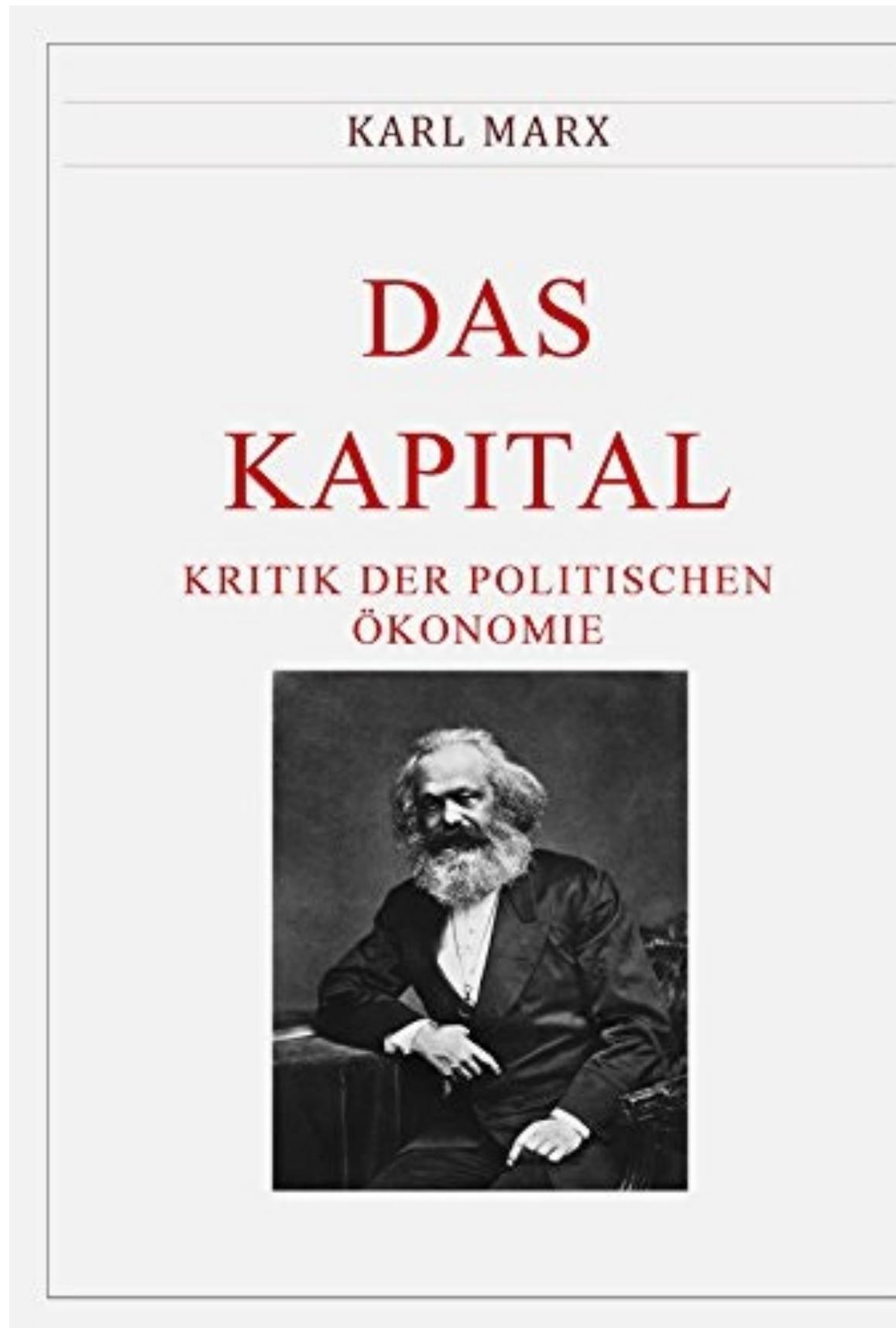
1x Compute



32x Compute

# Humanitarian Issues

- Democratization
- Environment & Risk



# About this course

# Coverage

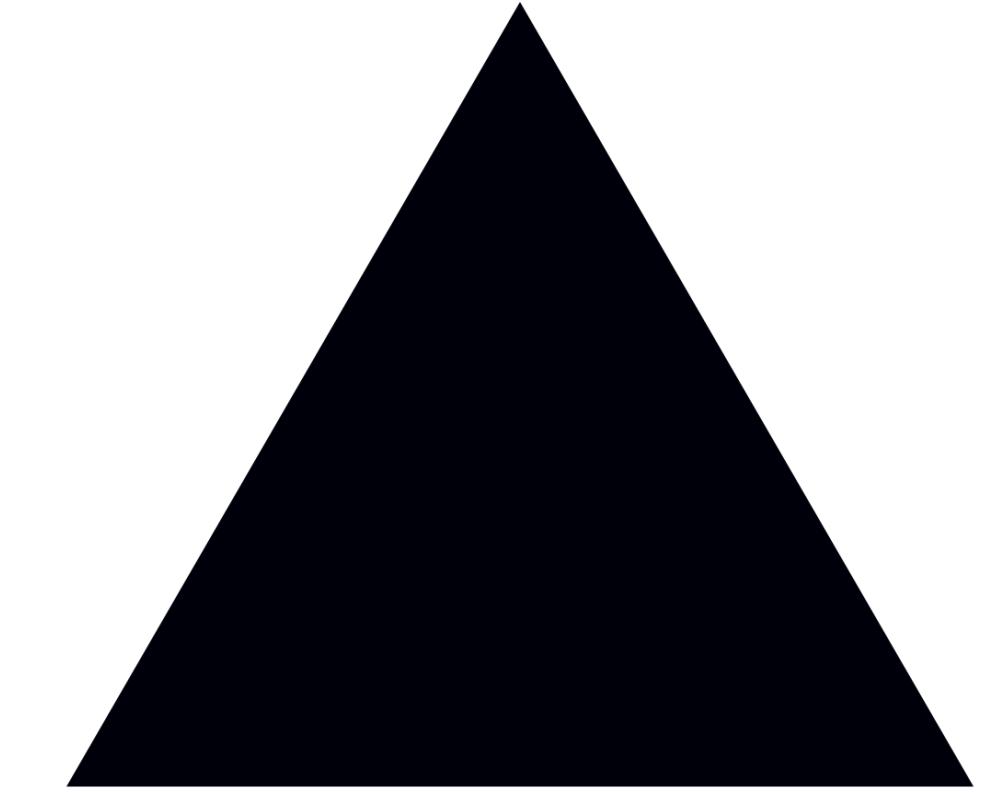
- Efficient AI, from the algorithmic side
- Goal. Help you confidently answer...
  - How much computation & memory will my model need?
  - How can we make my model run faster?
  - How can we make my model be lighter?
  - How can I reduce the training cost?

# Ultimately...

- Ultimately, we want to understand the **grand tradeoff**:

- Prediction Quality
- Training Cost
- Inference Cost

e.g., linear model is cheaper to train(?), cheaper to predict, but achieves lower performance than deep, overparameterized neural networks



- But our knowledge is not quite there yet ;(
- So let's be a bit more practical and applied!

# Schedule

## Phase 1. Lecture Only

- **W1.** Warm-Up
- **W2.** Sparsity
- **W3.** Quantization
- **W4.** NAS & KD
- **W5.** Efficient Training & Tuning
- **W6.** Adaptation
- **W7.** (No class; EU-KR AI meetings)
- **W8.** Parallelism

## Phase 2. Lecture + Student Presentation

- **W9.** Data Efficiency
- **W10.** LLM Compression
- **W11.** Long-Context LLMs
- **W12.** Low-Precision Training
- **W13.** Test-time Scaling
- **W14.** Efficient Diffusion Model
- **W15.** Efficient Neural Rendering
- **W16.** Final Presentation

# Prerequisites

- **Necessary**
  - Familiar with **deep learning**
    - Know how to code in Python
    - Experiences with PyTorch / TensorFlow / Jax
  - Have read 10+ academic papers
- **Recommended**
  - Knows how CPU / GPU / memory works

# Administrivia

# Team

- **Instructor.** Jaeho Lee [jaeho.lee@postech.ac.kr](mailto:jaeho.lee@postech.ac.kr)
  - Assistant Professor @ POSTECH EE
  - Visiting Researcher @ Google
  - Lectures
- **Teaching Assistant.** Hyunjong Ok [hyunjong.ok@postech.ac.kr](mailto:hyunjong.ok@postech.ac.kr)
  - M.S./Ph.D. Candidate @ POSTECH GSAI
  - Much expertise on Language & Speech
  - Assignments & Attendance (but please cc me as well)

# Hours & Location

- **Lectures**
  - PIAI 122
  - MW 11:00–12:15
- **Office Hours**
  - Terarosa Coffee
  - W 17:00–18:00 + by appointment
  - Tip. Think of it as a friendly coffee chat 

# Resources

- **Textbook.** None
  - Slides uploaded at [jaeho-lee.github.io](https://jaeho-lee.github.io)
    - some references there as well

# Grading

• Attendance	10%
• In-class Presentation	30%
• Project – Proposal	20%
• Project – Final	40%
<hr/>	
100%	

- **Note.** You get an F, if you (1) miss 3+ classes  
(2) use chatbots for assignments  
(3) cheat

# In-Class Presentation

- From W9–W15, we do
  - **Mon.** Lecture
  - **Wed.** Student Presentation
- By W2. I'll give you a paper list.
  - 3 papers for each week
  - By W3, You sign up as a presenter for the paper
    - Maximum 2 per paper



# In-Class Presentation

- At your presentation week, give a 20min talk, 5min Q&A.
- **Rubrics**
  - How clearly did you describe the **research question**?
  - How well did you describe the **proposed solution**?
  - How well did you identify the **limitations**  
(esp. the ones not mentioned in the paper)
  - How much did you **engage** in others' presentations?

# Project

- You'll prepare a submission to the **ICLR blog post** track, for ICLR 2026

- <https://iclr-blogposts.github.io/2025/about/>



**ICLR**

- **Idea**

1. Reviews past work and summarize the outcomes, develop new intuitions, or highlight some shortcomings.
2. Presents novel perspectives or interpretations of existing machine learning concepts or techniques.
3. Discusses important issues in machine learning, such as reproducibility, from a novel perspective.
4. Analyzes the societal implications of recent advancements in machine learning and AI.
5. Showcases cool research ideas that you tried but did not work out.

- Plus, make it about **Efficient ML**

# Project

- **Cool Examples**

- A Deeper Look at Zero-Cost Proxies for Lightweight NAS

<https://iclr-blog-track.github.io/2022/03/25/zero-cost-proxies/>

- Unraveling the Impact of Training Samples

<https://iclr-blogposts.github.io/2024/blog/unraveling-the-impact-of-training-samples/>

- Understanding In-Context Learning in Transformers

<https://iclr-blogposts.github.io/2024/blog/understanding-icl/>

# Project

- Max 3 people per group.
- **Proposal (W7)**
  - Write a one-page description of what you'll do. LaTeXed.
- **Final Poster Session (W16)**
  - By W15, submit your blog post via PLMS
  - By W15, Prepare a (small) poster, in pdf
  - I'll print 'em out and get some 
  - We do a small poster session of our own

# Project

- Rubrics
  - Clarity
  - Soundness
  - Originality
  - TA's assessments
  - Peer assessments

That's it for today

