

# Deep Revolution

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- 2024 Nobel prizes in physics (deep learning) & chemistry (Google DeepMind) shook the scientific world, heralding the new era of AI-enabled science  
<https://www.nobelprize.org/prizes/physics/2024>  
<https://www.nobelprize.org/prizes/chemistry/2024>
- In January 2025, DeepSeek sent a shock wave to Wall Street, White House, & Silicon Valley

## AI stocks plunge as China's DeepSeek sends shock wave through Wall Street

A Chinese AI company called DeepSeek is sending a shock wave through Wall Street. Margaret Brennan explains.

**©CBS NEWS** 1/28/2025

Trump calls DeepSeek a 'wake-up call' for U.S. tech and welcomes China's AI gains **FORTUNE** 1/28/2025

Meta is reportedly scrambling 'war rooms' of engineers to figure out how DeepSeek's AI is beating everyone else at a fraction of the price **FORTUNE** 1/27/2025

# Key Computational Enablers of DeepSeek?

- DeepSeek is a large language model (LLM) that outperforms OpenAI's ChatGPT with less computing
- **Multi-head Latent Attention** guarantees efficient inference through significantly compressing the Key-Value cache into a latent vector, while **DeepSeekMoE (Mixture-of-Experts)** enables training strong models at an economical cost through **sparse computation**

<https://arxiv.org/abs/2405.04434>

- DeepSeek-V3 pioneers an **auxiliary-loss-free strategy for load balancing** and sets a **multi-token prediction training objective for stronger performance**

<https://arxiv.org/html/2412.19437v1>

- Will brain-like sparse spiking of neurons solve the AI power catastrophe (*cf.* Google's Pathways)?

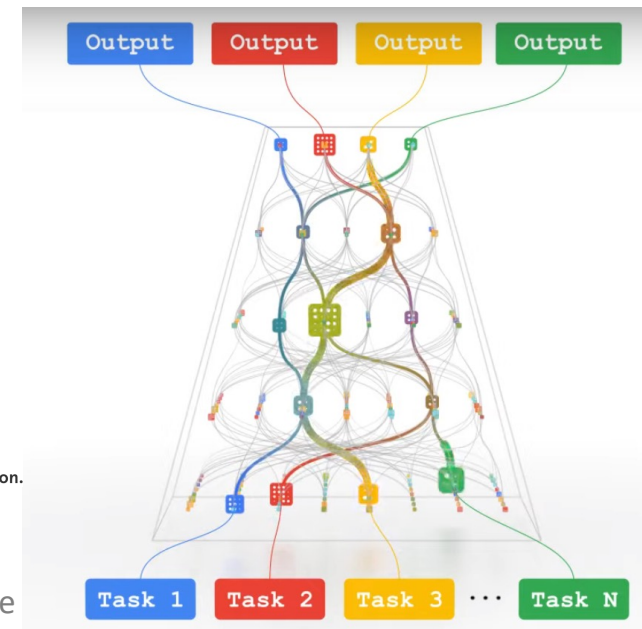
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WIRED

## AI's Energy Demands Are Out of Control. Welcome to the Internet's Hyper-Consumption Era

Generative artificial intelligence tools, now part of the everyday user experience online, are causing stress on local power grids and mass water evaporation.

<https://blog.google/technology/ai/introducing-pathways-next-generation-ai-architecture>

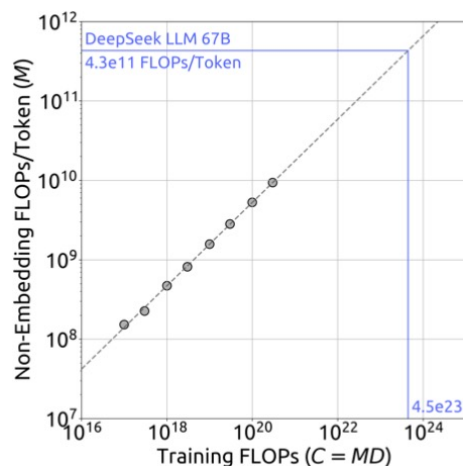


# Scaling Analysis Is Important

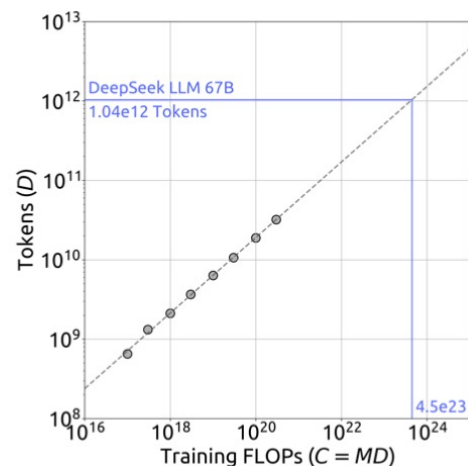
- Understanding scaling laws of LLMs is essential for long-term projection

<https://arxiv.org/abs/2401.02954>

- Use the same scaling exponent analysis (log-log plot & linear fit) as in assignment 2, Part I-2!



(b) Optimal model scaling



(c) Optimal data scaling

$$M_{\text{opt}} = M_{\text{base}} \cdot C^a, \quad M_{\text{base}} = 0.1715, \quad a = 0.5243$$
$$D_{\text{opt}} = D_{\text{base}} \cdot C^b, \quad D_{\text{base}} = 5.8316, \quad b = 0.4757$$

Approach	Coeff. $a$ where $N_{\text{opt}}(M_{\text{opt}}) \propto C^a$	Coeff. $b$ where $D_{\text{opt}} \propto C^b$
OpenAI (OpenWebText2)	0.73	0.27
Chinchilla (MassiveText)	0.49	0.51
Ours (Early Data)	0.450	0.550
Ours (Current Data)	0.524	0.476
Ours (OpenWebText2)	0.578	0.422