

# Fundamentals of AI/ML: From Turing Test to Transformers

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# What is Artificial Intelligence?

**Definition:** The ability of machines to perform tasks that typically require human intelligence

## Key Components:

- ▶ Learning from experience
- ▶ Reasoning and problem-solving
- ▶ Understanding language
- ▶ Perception and recognition
- ▶ Planning and decision-making

**Fun Fact:** The term "Artificial Intelligence" was coined just to avoid using "cybernetics" because the researchers didn't want Norbert Wiener as their guru!

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## **Wartime Contributions:**

- ▶ Enigma code breaking at Bletchley Park
- ▶ Shortened WWII by an estimated 2–4 years

# The Birth of AI - Dartmouth Conference (1956)

## The Constitutional Convention of AI– Organizers:

- ▶ John McCarthy (coined "Artificial Intelligence"; pioneer of Lisp and AI research)
- ▶ Marvin Minsky (cognitive scientist; co-founder of MIT AI Lab)
- ▶ Claude Shannon (father of information theory)
- ▶ Nathaniel Rochester (IBM engineer; designer of the IBM 701)

## Goals:

- ▶ "Every aspect of learning can be precisely described so that a machine can simulate it"
- ▶ 6-8 week brainstorming session
- ▶ Foundation of AI as a research field

## Outcome:

Birth of AI optimism - "Machines will be capable of doing any work a man can do" within 20 years

# Early AI Achievements (1950s-1960s)

## **ELIZA (1966):**

- ▶ First chatbot by Joseph Weizenbaum
- ▶ Simulated psychotherapist
- ▶ Demonstrated how simple pattern matching could fool users

## **Logic Theorist (1956):**

- ▶ First AI program
- ▶ Proved mathematical theorems
- ▶ Written during Dartmouth Conference

## **General Problem Solver (1957):**

- ▶ Herbert Simon and Allen Newell
- ▶ First general-purpose problem-solving program



# The First AI Winter (1974-1980)

## What Went Wrong?

- ▶ Overpromising and underdelivering
- ▶ Limited computational power
- ▶ Lack of real-world applications
- ▶ The Lighthill Report (1973) criticized AI research

## Key Problems:

- ▶ Combinatorial explosion
- ▶ Frame problem
- ▶ Common sense reasoning
- ▶ Limited memory and processing power

## Funding Impact:

- ▶ DARPA cut AI funding
- ▶ British government reduced support
- ▶ Academic positions disappeared

**Lesson:** Hype cycles are dangerous in AI!

# The Expert Systems Boom (1980-1987)

## What are Expert Systems?

- ▶ Programs that capture expert knowledge in specific domains
- ▶ Rule-based reasoning
- ▶ If-then logic structures

## Success Stories:

- ▶ **MYCIN (1972)**: Medical diagnosis
- ▶ **XCON/R1 (1980)**: Saved DEC \$40 million annually
- ▶ **DENDRAL (1965)**: Chemical analysis

## Why They Worked:

- ▶ Narrow, well-defined domains
- ▶ Clear rules and logic
- ▶ Practical business value

# The Second AI Winter (1987-1993)

## **The Expert Systems Bubble Burst:**

- ▶ Expensive to maintain
- ▶ Brittle and inflexible
- ▶ Couldn't handle uncertainty
- ▶ Knowledge acquisition bottleneck

## **Economic Factors:**

- ▶ LISP machine market collapsed
- ▶ Japan's Fifth Generation project failed
- ▶ Desktop PCs became more powerful than LISP machines

**Silver Lining:** This winter led to more realistic expectations and better methodologies

# Computing Revolution Enables AI Renaissance

## Moore's Law Impact:

- ▶ Computing power doubles every 2 years
- ▶ Enabled complex algorithms
- ▶ Made large-scale data processing possible

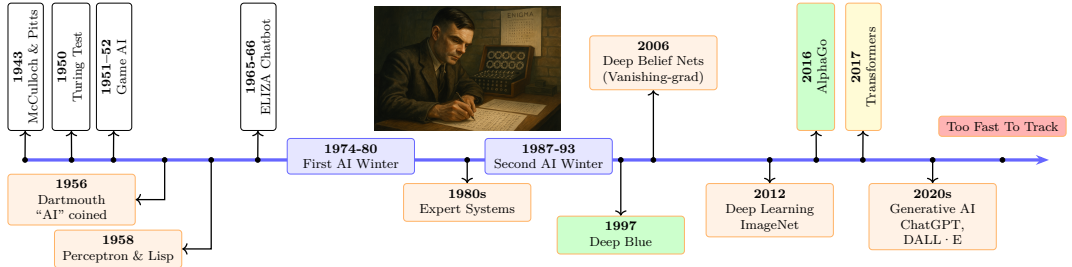
## GPU Revolution:

- ▶ Originally for graphics
- ▶ Perfect for parallel computation
- ▶ Neural networks need massive parallelization
- ▶ NVIDIA becomes AI kingmaker

## The Internet:

- ▶ Massive datasets available
- ▶ Distributed computing
- ▶ Real-world testing platforms

# Timeline of AI



# GPU vs CPU for AI

## CPU:

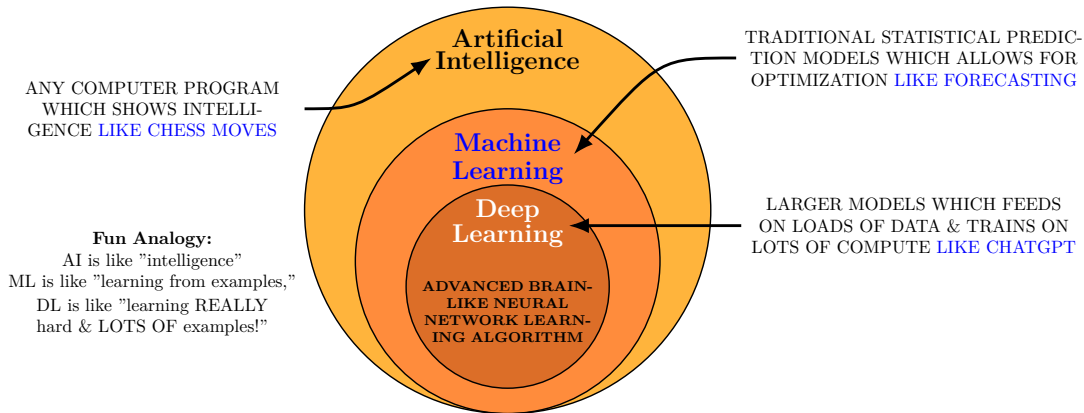
- ▶ Few powerful cores
- ▶ Sequential processing
- ▶ Complex instruction sets
- ▶ Good for general computing

## GPU:

- ▶ Thousands of simple cores
- ▶ Parallel processing
- ▶ Perfect for matrix operations
- ▶ 10x faster for neural networks

**Key Insight:** AI workloads are embarrassingly parallel - perfect for GPUs!

# AI, Machine Learning, and Deep Learning



# Machine Learning Types

	<b>Supervised Learning</b>	<b>Unsupervised Learning</b>	<b>Reinforcement Learning</b>
<b>Description</b>	Trained on Labeled Data	Trained on Unlabeled Data	Learn by Interacting with Environment
<b>Goal</b>	Make Predictions	Find Patterns and Structure	Maximize Reward
<b>Examples</b>	Spam Detection Image Classification	Clustering Dimensionality Reduction	Game Playing Robotics LLMs



# The Deep Learning Revolution (2006-2012)

## **Geoffrey Hinton's Comeback:**

- ▶ Coined "Deep Learning" in 2006
- ▶ Solved the vanishing gradient problem
- ▶ Layer-by-layer pre-training

## **AlexNet (2012):**

- ▶ Won ImageNet competition
- ▶ 84% → 15% error rate improvement
- ▶ Proved deep learning's potential

## **Key Enablers:**

- ▶ Big data (ImageNet, etc.)
- ▶ GPU computing
- ▶ Better algorithms (ReLU, dropout, batch normalization)

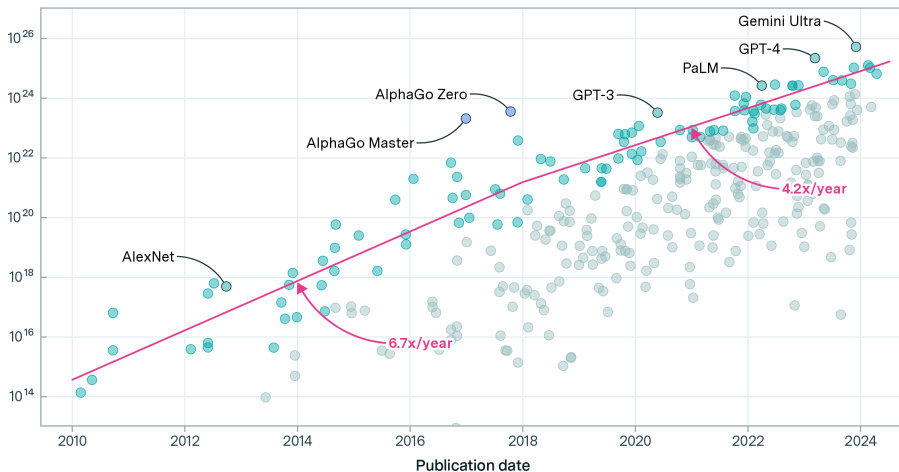
# AI Compute Growth

## Training compute of frontier models



Training compute (FLOP)

● Frontier ● Non-frontier ● Outliers 96 frontier models



# AI Today - Capabilities and Limitations

## What AI Can Do Well:

- ▶ Pattern recognition
- ▶ Language translation
- ▶ Code generation
- ▶ Creative writing
- ▶ Game playing
- ▶ Image generation

## What AI Still Struggles With:

- ▶ Common sense reasoning
- ▶ Causal understanding
- ▶ Long-term planning
- ▶ Robustness and safety
- ▶ Understanding vs. pattern matching

# Future Directions and Challenges

## Technical Challenges:

- ▶ Artificial General Intelligence (AGI)
- ▶ Explainable AI
- ▶ Few-shot learning
- ▶ Multimodal understanding
- ▶ Energy efficiency

## Societal Challenges:

- ▶ Job displacement
- ▶ Bias and fairness
- ▶ Privacy and surveillance
- ▶ Autonomous weapons
- ▶ Misinformation

AI For Engineering & Science

# Questions!