





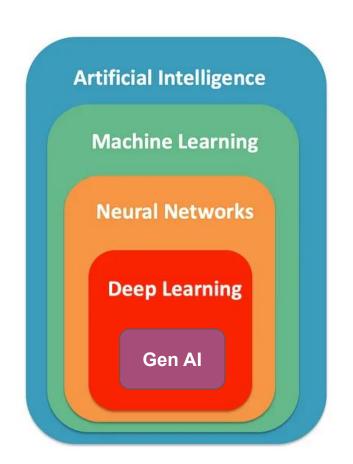
The capability of computational systems to perform tasks typically associated with human intelligence, such as learning, reasoning, problem-solving, perception, and decision-making.

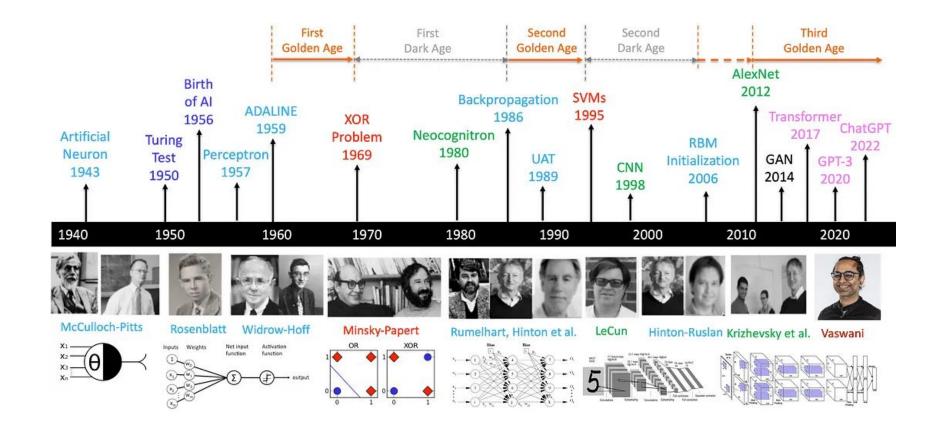


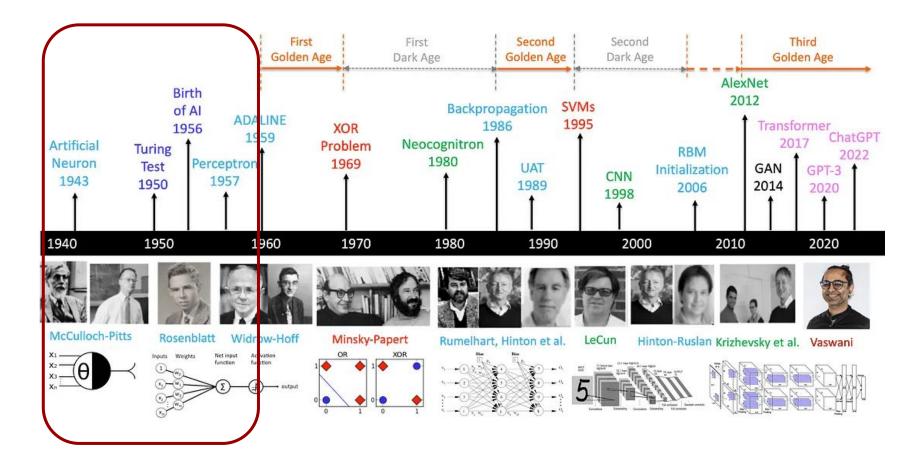
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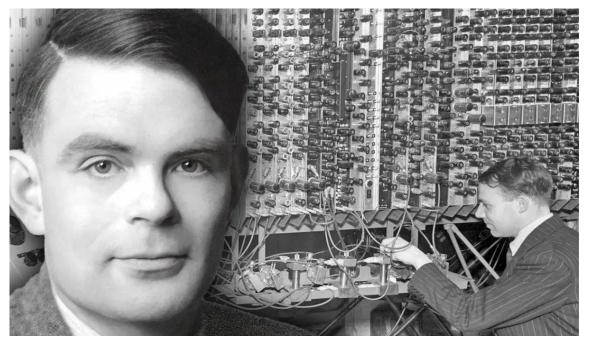
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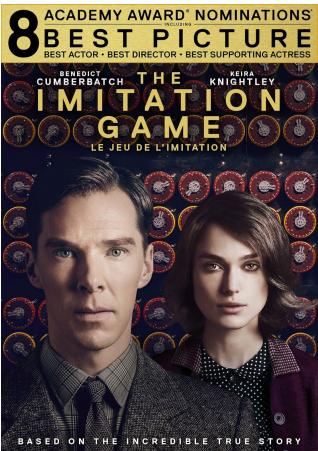




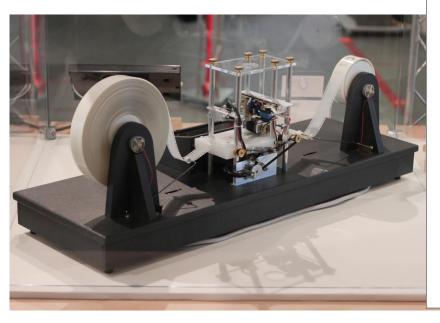


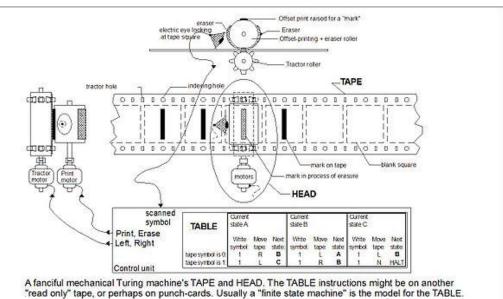
Alan Turing defines the universal computer, 1936





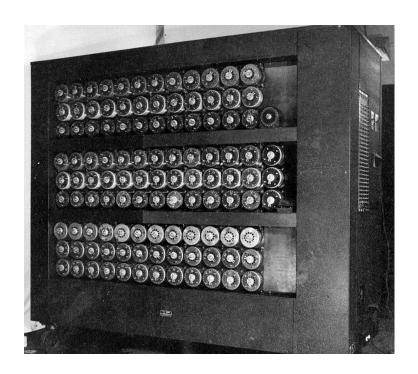
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  - Turing Machine



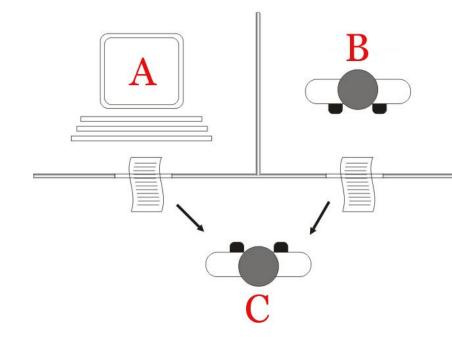


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  - Turing machine
  - Enigma & Bombe machine





- Alan Turing defines the universal computer, 1936
  - Turing machine
  - Enigma & Bombe machine
  - Turing test: Can a machine behave indistinguishably from a human?



### Today

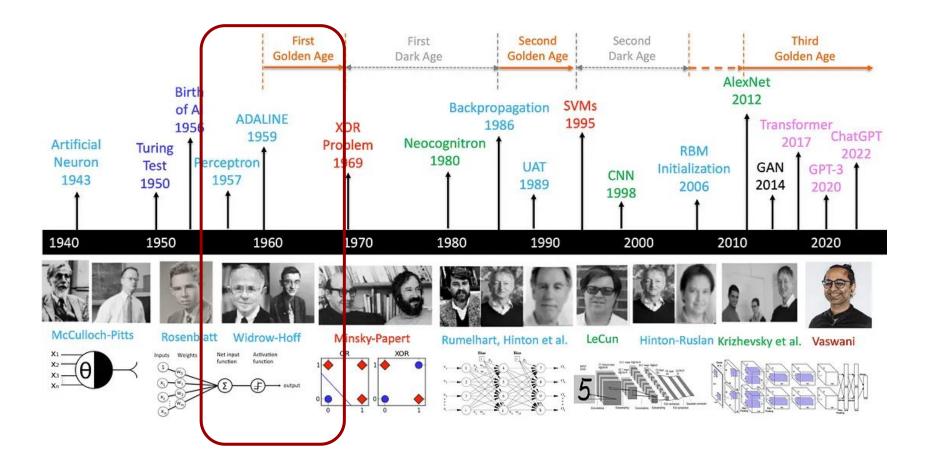


#### First half:

- Some logistics
- Finishing up history of Al

#### Second half:

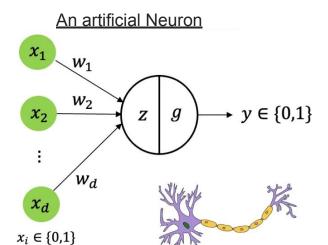
- Data Science tools (conceptual tools today, technical ones next week)
- Example: Customer churn problem
- Group Project

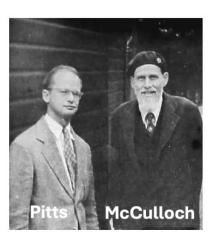


The Dartmouth conference marked the birth of AI, 1956



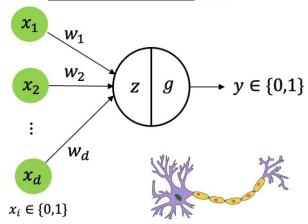
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#### An artificial Neuron



#### Both inputs and output are binary

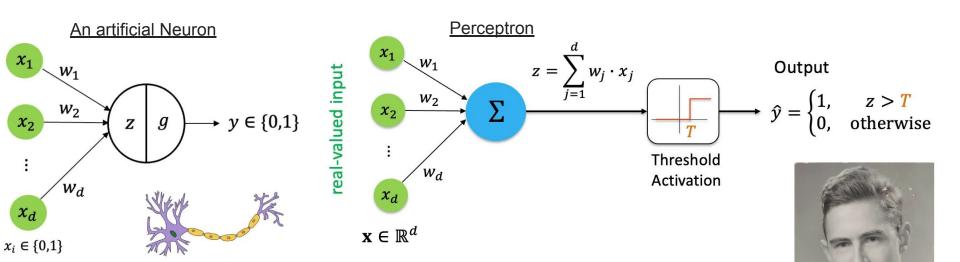
•  $x_i \in \{0,1\}$  and  $y \in \{0,1\}$ 

$$z(x_1, x_2, \dots, x_d) = z(\mathbf{x}) = \sum_{j=1}^d w_j \cdot x_j$$

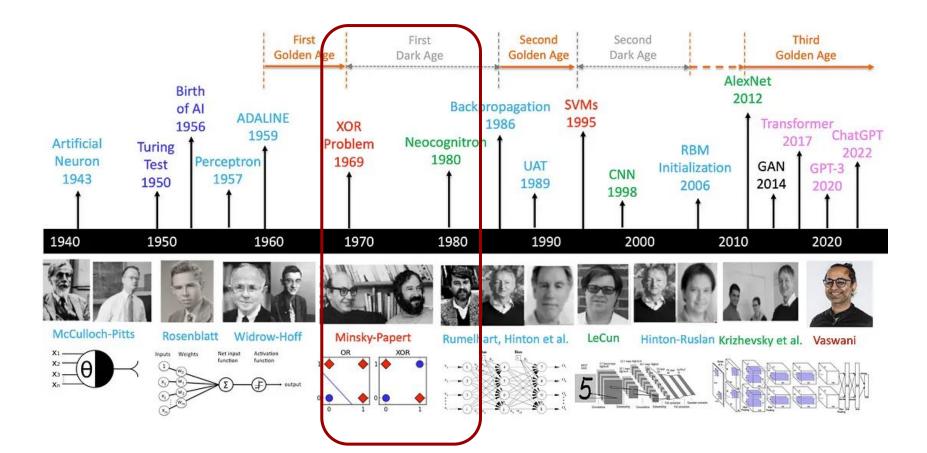
$$y = g(z(\mathbf{x})) = \begin{cases} 1 & \text{if } z(\mathbf{x}) \ge T \\ 0 & \text{if } z(\mathbf{x}) < T \end{cases}$$

Threshold Activation function

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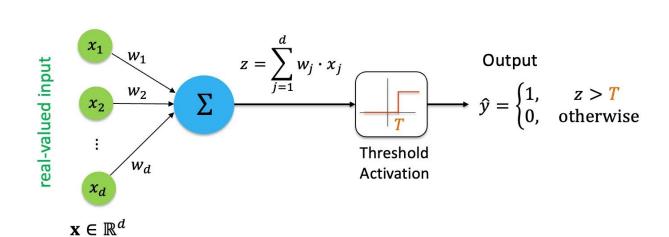


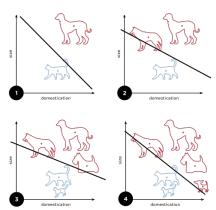
**Frank Rosenblatt** 

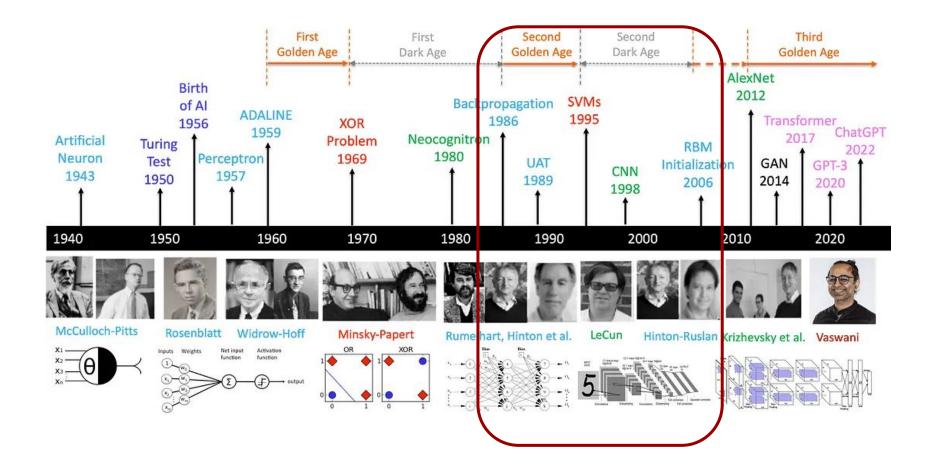


### AI Winter 1974-1990

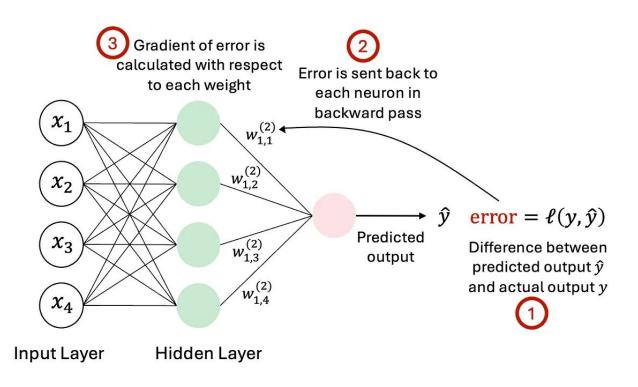
- 1965: Dreyfus, "What computers can't do"
- 1969: Minsky and Papert publish Perceptrons
- 1974: DARPA cuts AI funding
- 1980: Searle's Chinese Room Argument, in "Minds, brains, and programs"



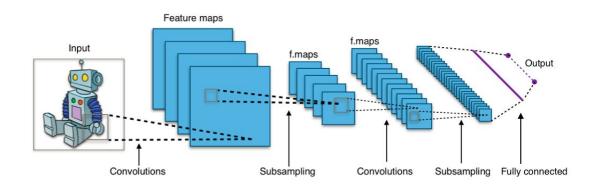




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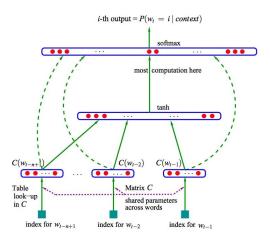
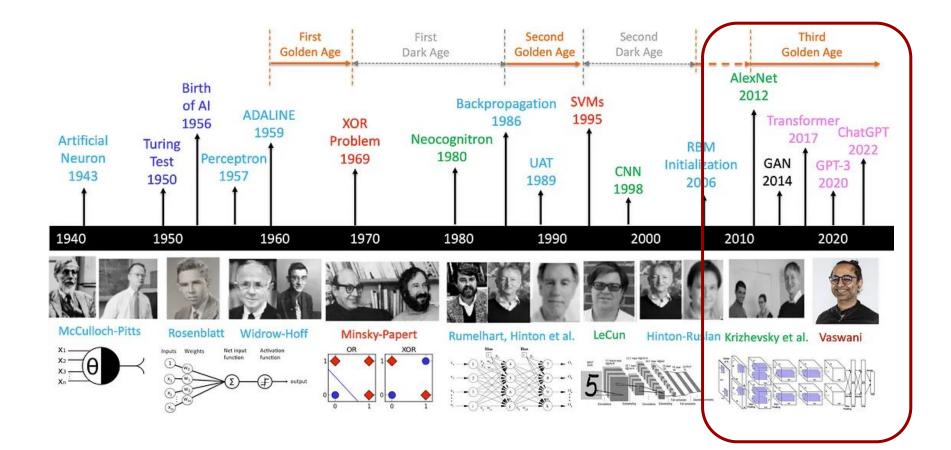


Figure 1: Neural architecture:  $f(i, w_{t-1}, \dots, w_{t-n+1}) = g(i, C(w_{t-1}), \dots, C(w_{t-n+1}))$  where g is the neural network and C(i) is the i-th word feature vector.

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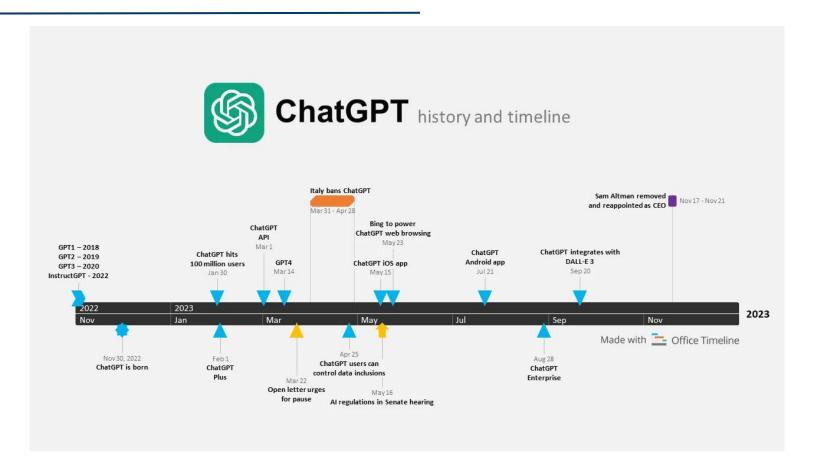






### Golden age again!

- 2012: Hinton uses Deep ConvNets to solve ImageNet classification
- 2013: Hinton hired by Google, uses NN on mobile
- Boston Dynamics acquired by Google
- Yann LeCun heads FAIR for Facebook
- 2014: Google acquires DeepMind
- 2015: Musk donates \$10m to Future of Life to study existential risk for AI
- Autonomous vehicle testing begins
- 2016: AlphaGo beats Lee Sedol
- 2017: AlphaZero beats beats Stockfish



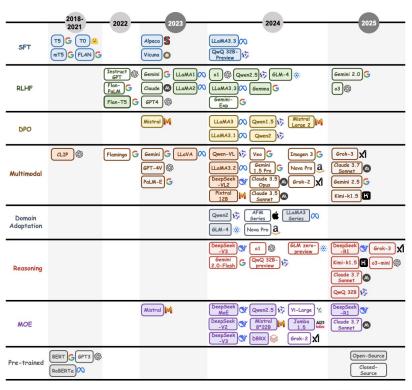
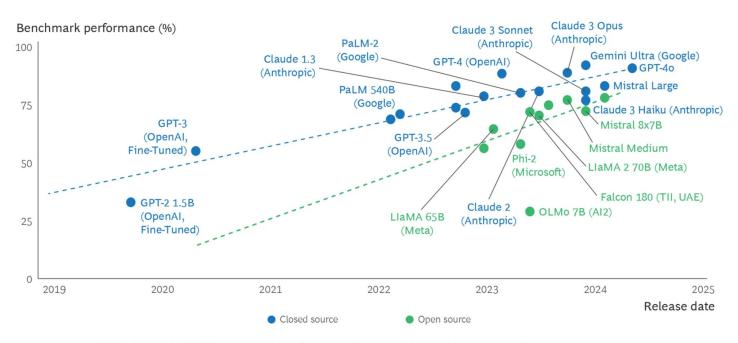


Figure 3: Timeline of post-training technique development for Large Language Models (2018–2025), delineating key milestones in their historical progression.

#### Exhibit 2 - The Performance Curve of GenAI Model Evolution



Sources: Open LLM leaderboard; HELM leaderboard; Chatbot Arena leaderboard; expert interviews; BCG analysis.

**Note:** Performance is calculated with the Measuring Massive Multitask Language Understanding (MMLU) benchmark, one of several tests used to estimate model performance and accuracy. Data is current as of May 2024.

