

# Introduction to Artificial Intelligence

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**Cyber Crisis Management:  
Navigating Disinformation and  
Cyber Attacks in the AI Era**



Federal Foreign Office



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# Module Outline

- Terminology and History
- Neural Networks
- The Generative AI Wave
- Modeling Language and Imagery
- Challenges and the Future of AI

Artificial Intelligence

Machine Learning

Neural Networks

Deep Learning

Generative AI

LLM

# Definitions

## ■ Artificial Intelligence

- *The capability for machines to imitate intelligence human behavior*

## ■ Machine Learning

- *An application of AI that allows computer systems to automatically learn from experience without explicit programming*

## ■ Neural Networks

- *An attempt to model networks of neurons in the brain with computational circuits*

## ■ Deep Learning

- *An application of neural networks strengths organized into many layers*

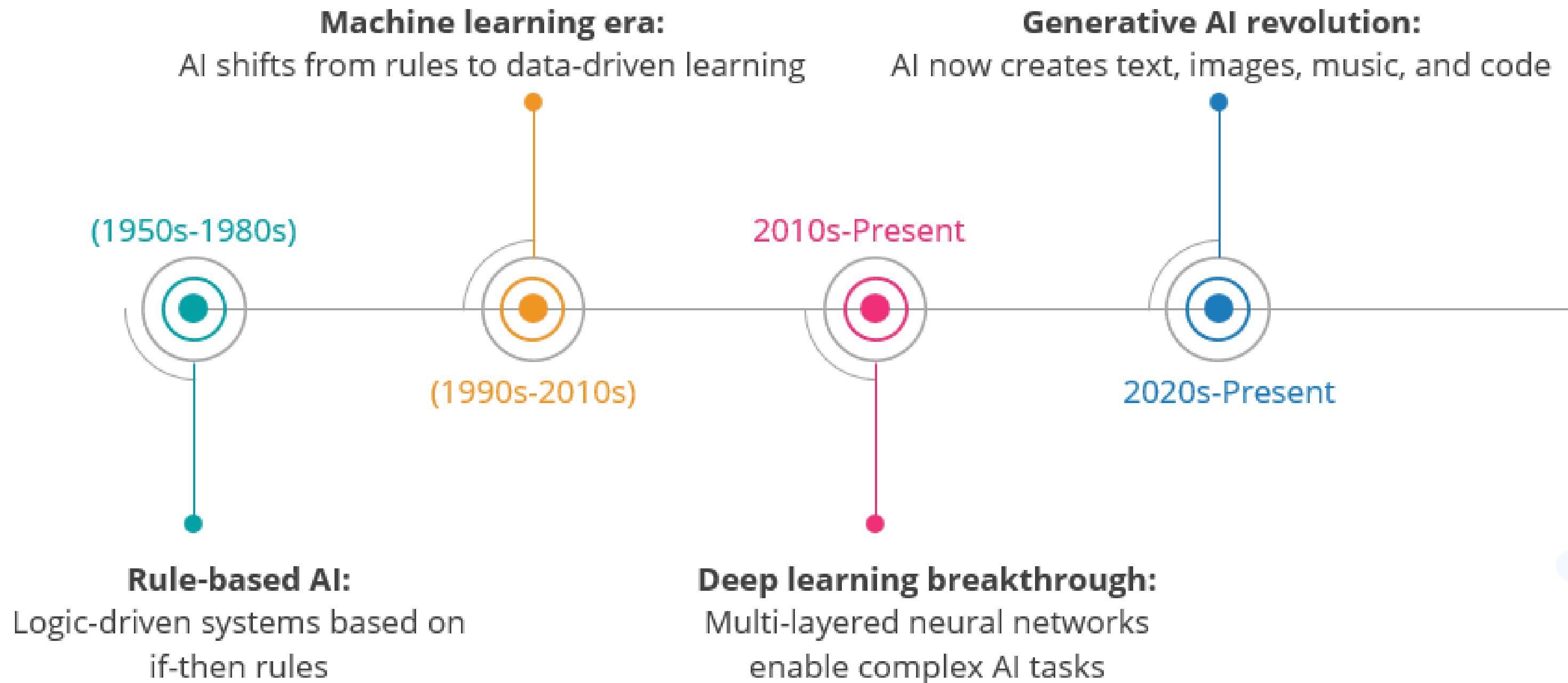
## ■ Generative AI

- *A type of deep learning that can create new content, including text, images and audio, in response to a query or prompt*

## ■ Large Language Model (LLM)

- *A type of generative AI that can perform a variety of natural language processing (NLP) tasks, including generating and classifying text, answering questions in a conversational manner and translating text from one language to another*

# How Did We Get Here?

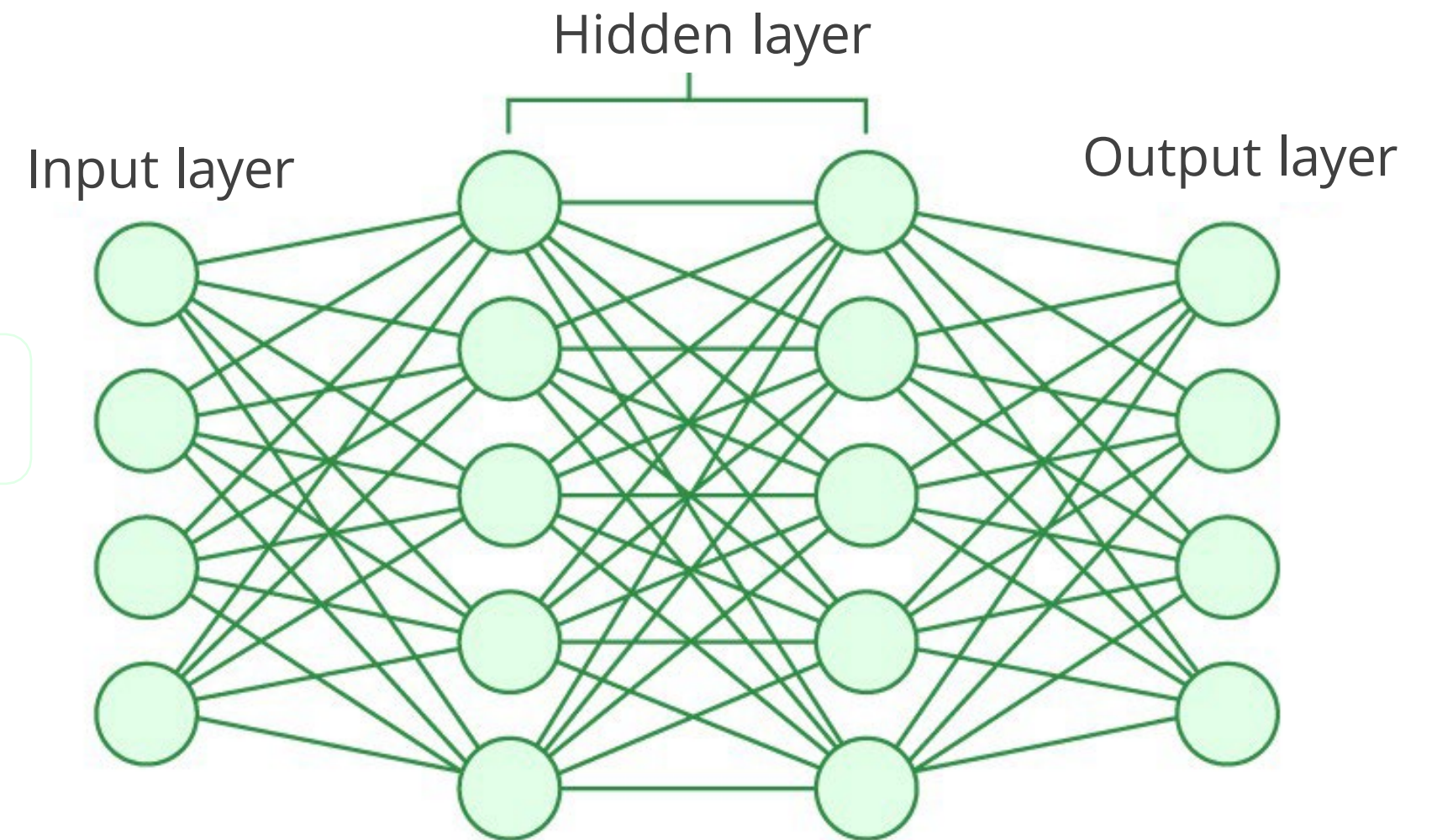


Courtesy of Purdue University



# Neural Networks

- Mimics the human brain
- The work horse of deep learning
- Most common types are recurrent neural networks and convolutional neural networks

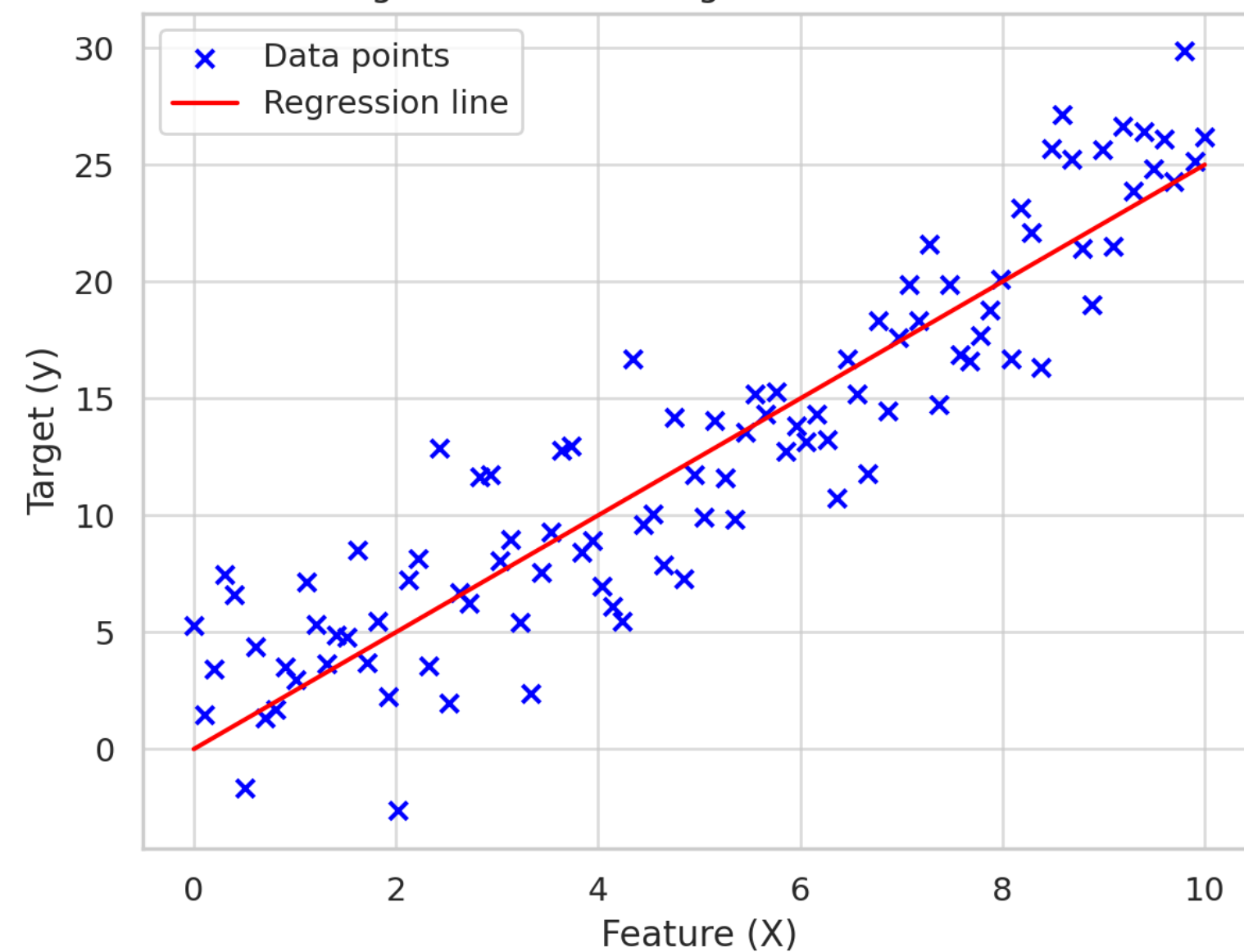


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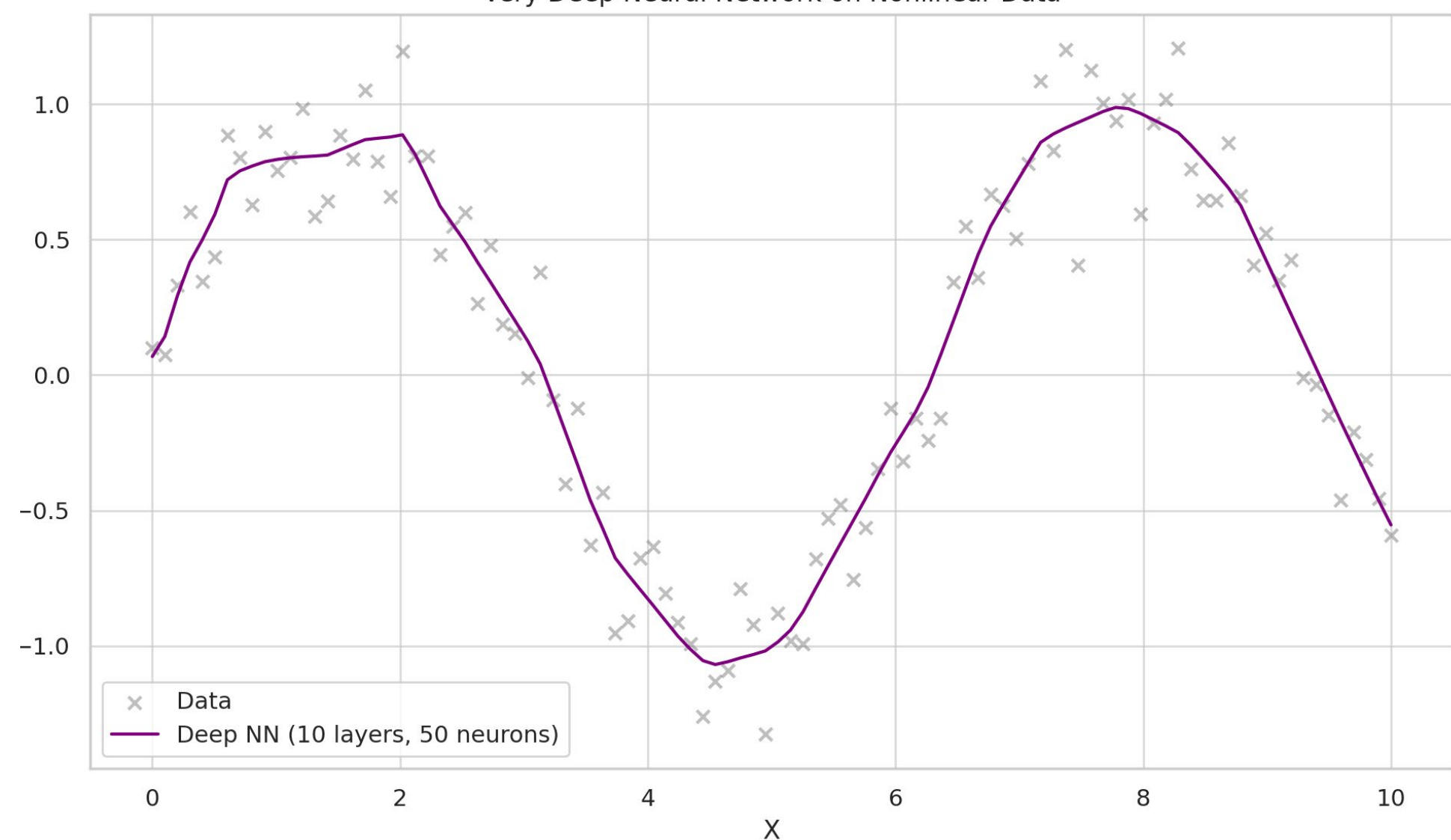


# Regression

Regression: Predicting a Continuous Value

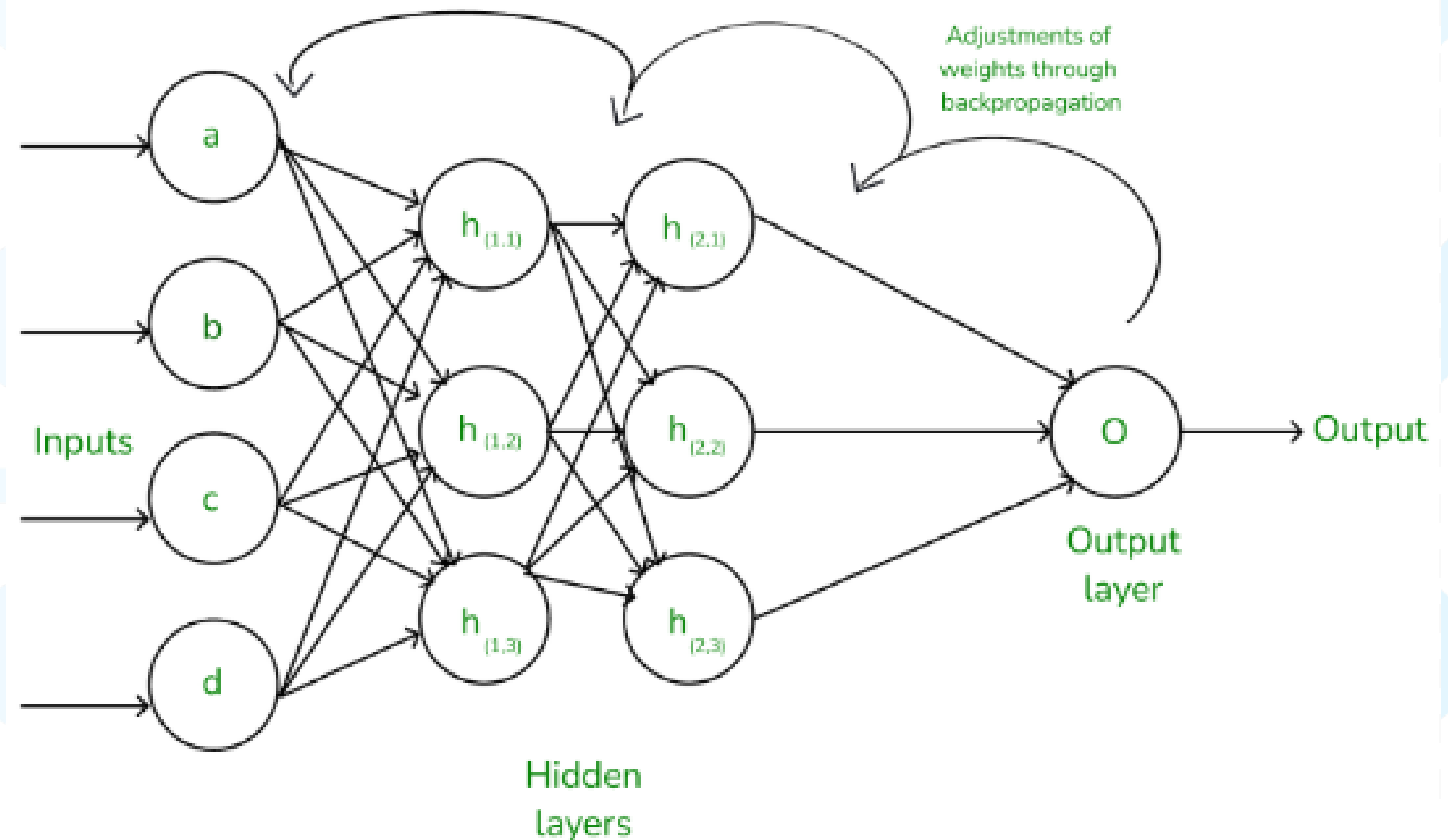


Very Deep Neural Network on Nonlinear Data



# Back Propagation

- We start with a set of inputs
- We make a forward pass through the network to get an output
- We calculate the “loss” or difference between the actual and the expected output
- Then use back propagation to go back through the network adjusting the weights to minimize the loss
- Rinse and repeat till we get the right answer





# The Generative AI Wave

- Massive Compute Power
- Big Data
- Transformer Architecture

## Attention Is All You Need

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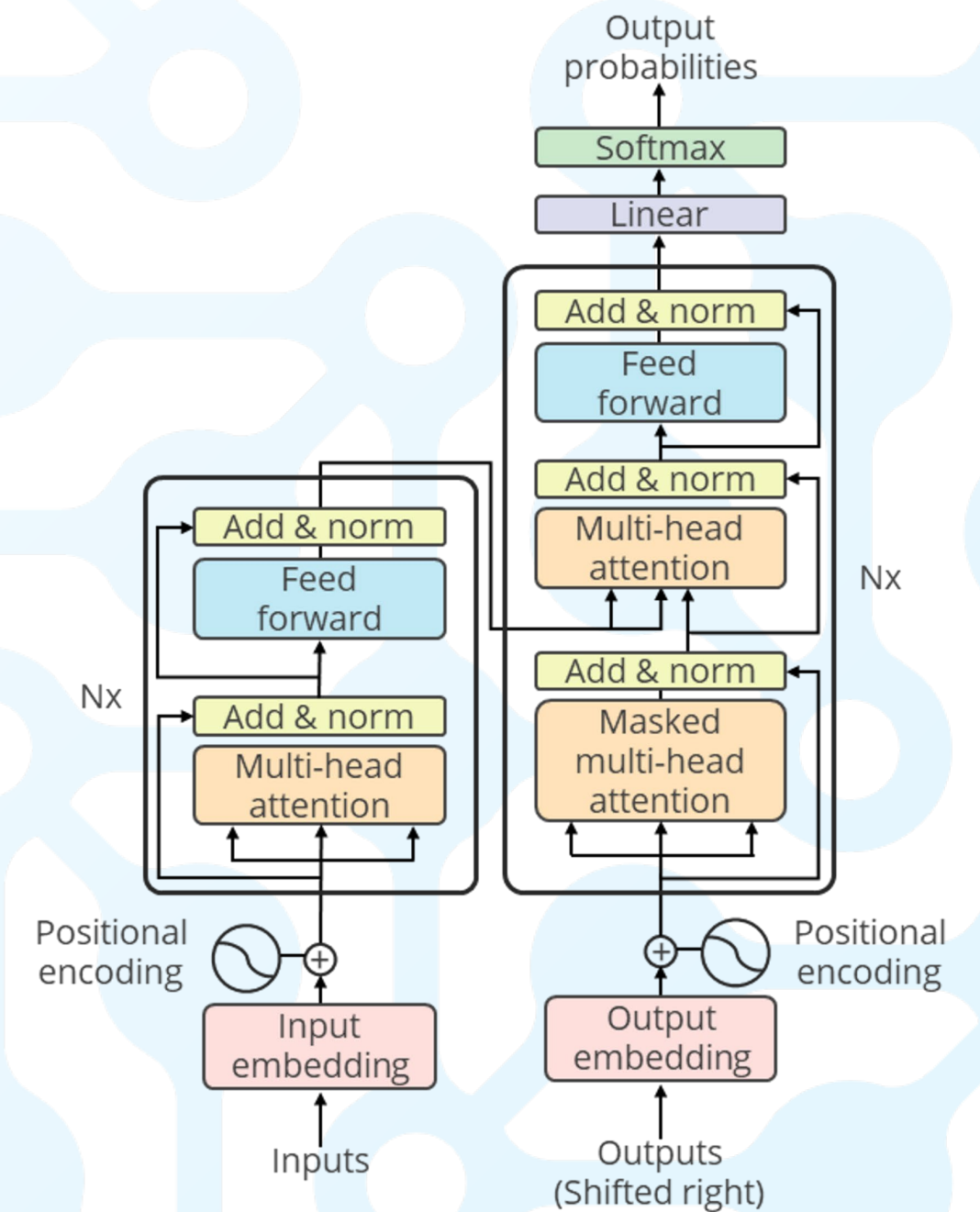
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**Abstract**

The dominant sequence transduction models are based on complex recurrent or convolutional neural networks that include an encoder and a decoder. The best performing models also connect the encoder and decoder through an attention mechanism. We propose a new simple network architecture, the Transformer, based solely on attention mechanisms, dispensing with recurrence and convolutions entirely. Experiments on two machine translation tasks show these models to be superior in quality while being more parallelizable and requiring significantly less time to train. Our model achieves 28.4 BLEU on the WMT 2014 English-to-German translation task, improving over the existing best results, including ensembles, by over 2 BLEU. On the WMT 2014 English-to-French translation task, our model establishes a new single-model state-of-the-art BLEU score of 41.8 after training for 3.5 days on eight GPUs, a small fraction of the training costs of the best models from the literature. We show that the Transformer generalizes well to other tasks by applying it successfully to English constituency parsing both with large and limited training data.



# Generative AI can be Creative

- Think of GenAI as a Chef
- with a set of ingredients (data) and recipes (algorithms)
- Instead of following a fixed recipe the chef might experiment, creating new dishes with surprising new flavors
- Similarly GenAI can use data and algorithms to cook up new content, adding its own creative touch





# Large Language Models



GPT 4



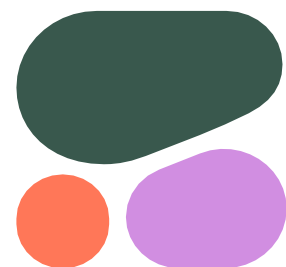
DeepSeek



Claude



BERT



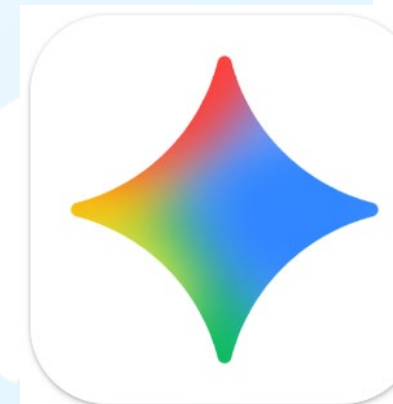
Cohere



Falcon



LLaMA



Gemini

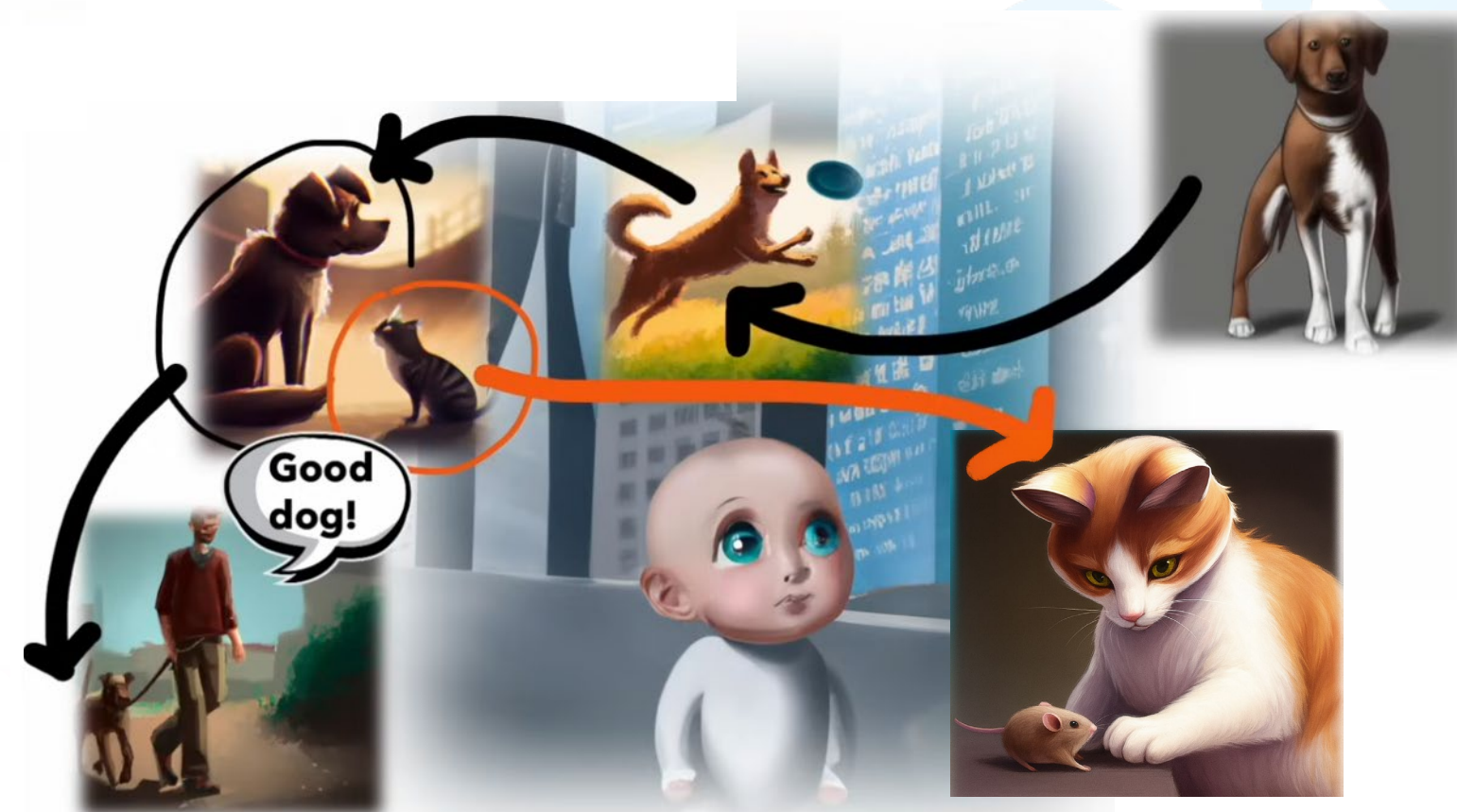


# Analogies to Training Approaches

## Supervised Learning



## Self-Supervised Learning

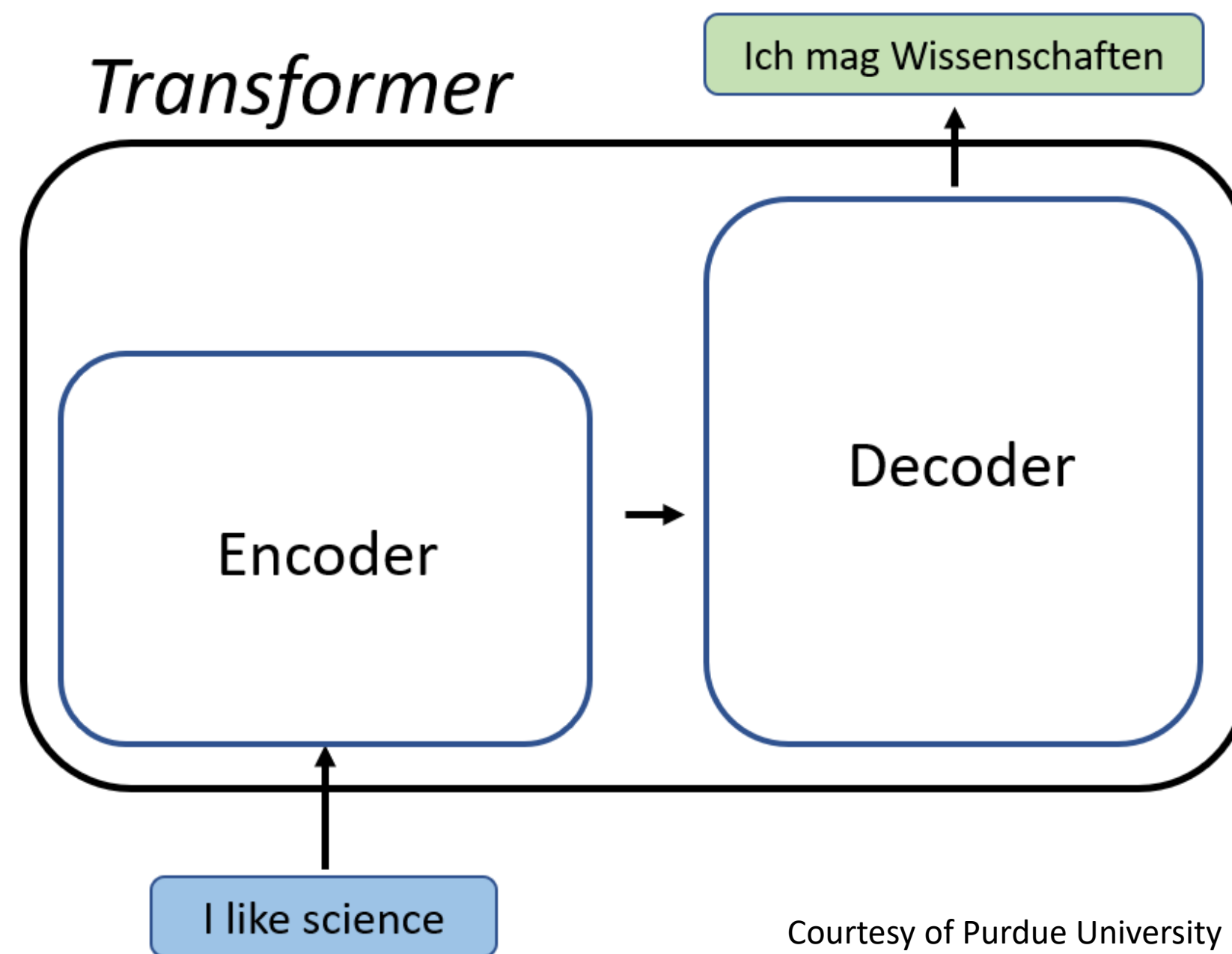


Courtesy of Rickard Br  l Gabrielsson, MIT

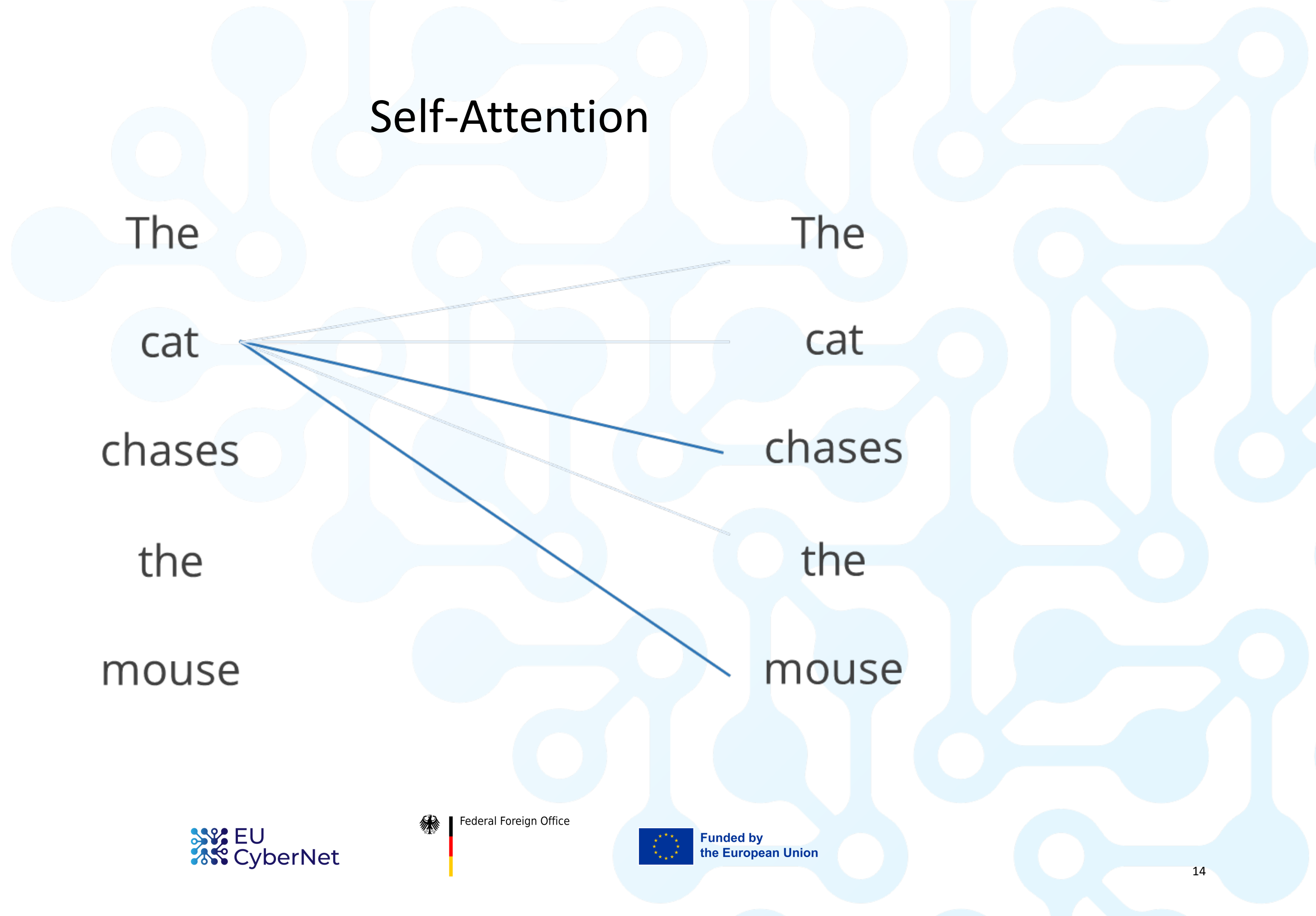




# The Transformer Architecture



Courtesy of Purdue University



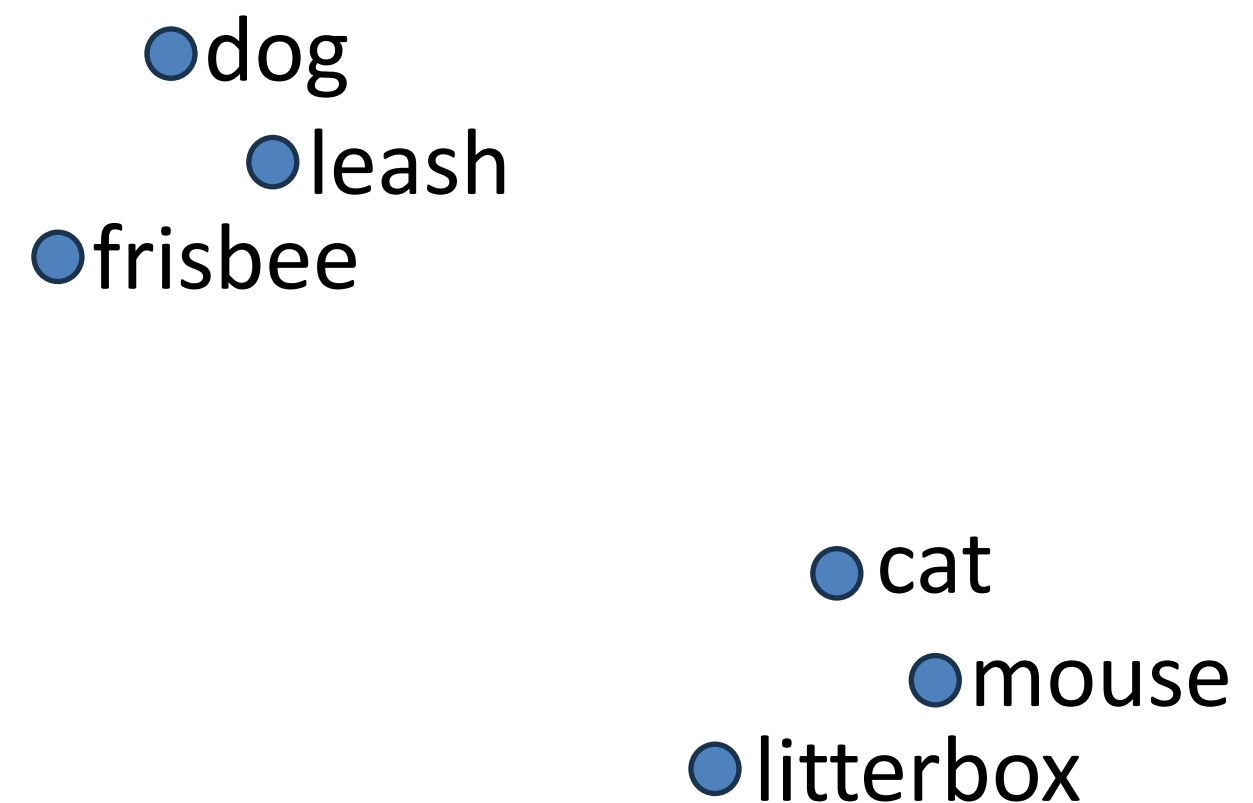
# Word Embeddings

- Words are represented as points in an n-dimensional space
- This allows us to capture the complexity and nuance of language in mathematical form
- We use tensor algebra to quantify similarity and relationships and to transform neural states

Word Coordinates			
	Gender	Age	
grandfather	[ 1,	9	]
man	[ 1,	7	]
adult	[ 5,	7	]
woman	[ 9,	7	]
boy	[ 1,	2	]
child	[ 5,	2	]
girl	[ 9,	2	]
infant	[ 5,	1	]

# Contextual Word Embeddings

- “You shall know a word by the company it keeps”
- A contextual representation maps both a word and the context into its embedding
- Words that often co-occur end up close to each other in the vector space







# Language Modeling

$$P(\text{She likes ice cream}) = P(\text{She}) * P(\text{likes} | \text{She}) * P(\text{ice} | \text{She likes}) * P(\text{cream} | \text{She likes ice})$$

$P(\alpha | \text{There she built a})$

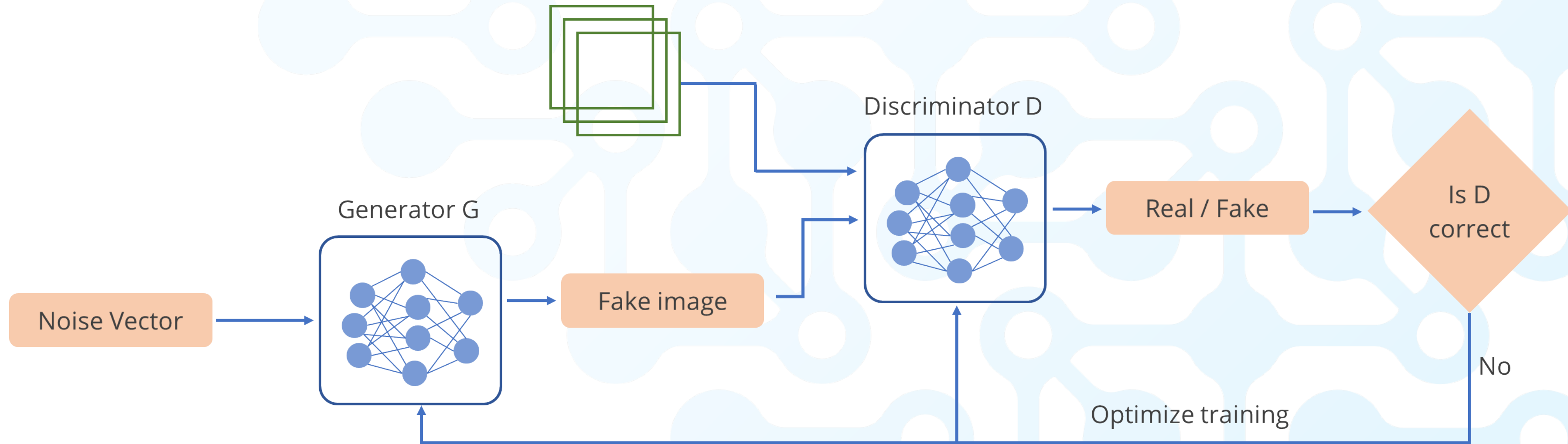
.....

$P(\alpha | \text{Alice went to the beach. There she built a})$

$P(\alpha | \text{Edward told me that Michael was very sick so I took him to the hospital})$

$P(\alpha | \text{Edward told me that Michael was very sick so I took him to see Michael})$

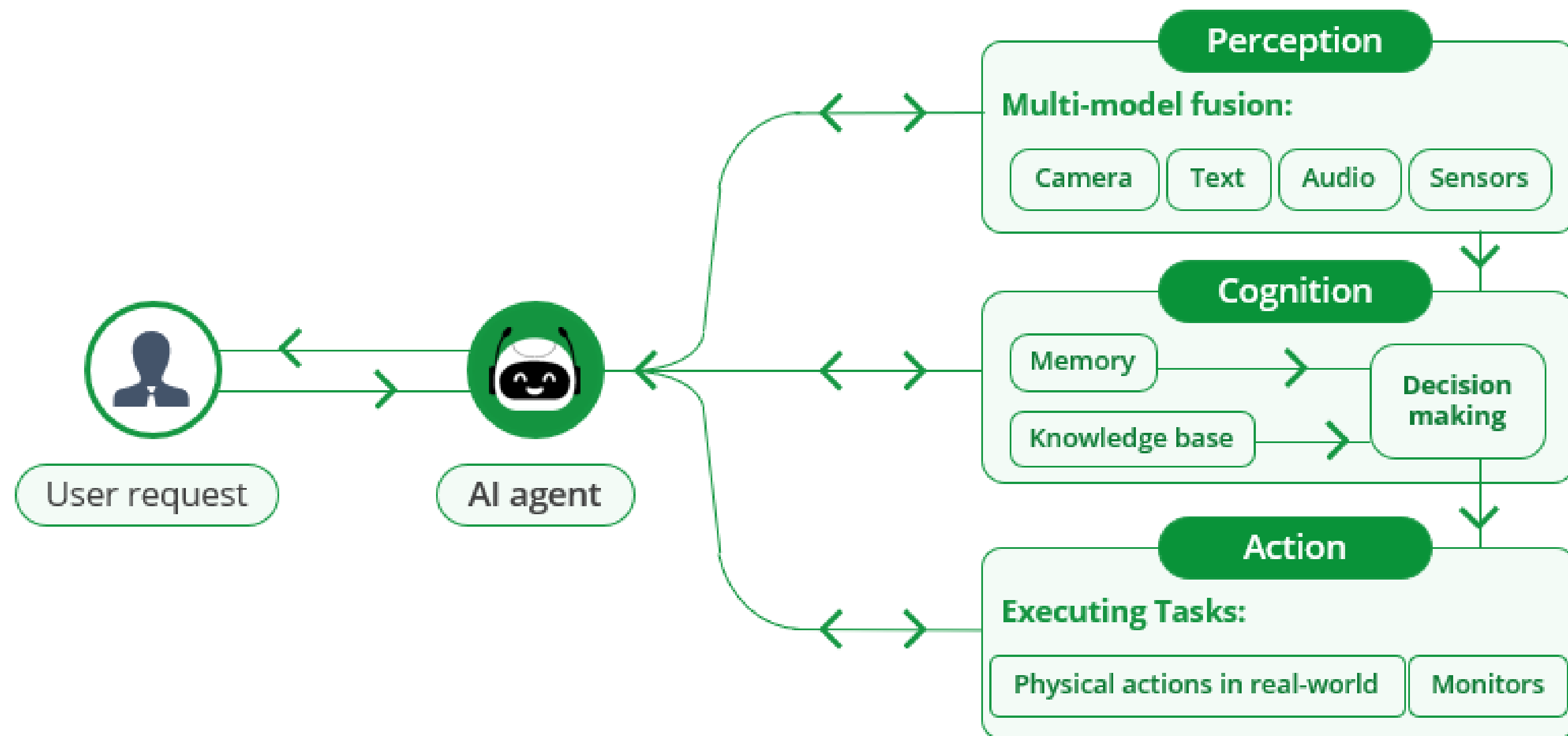
# Modeling Imagery with Generative Adversarial Networks



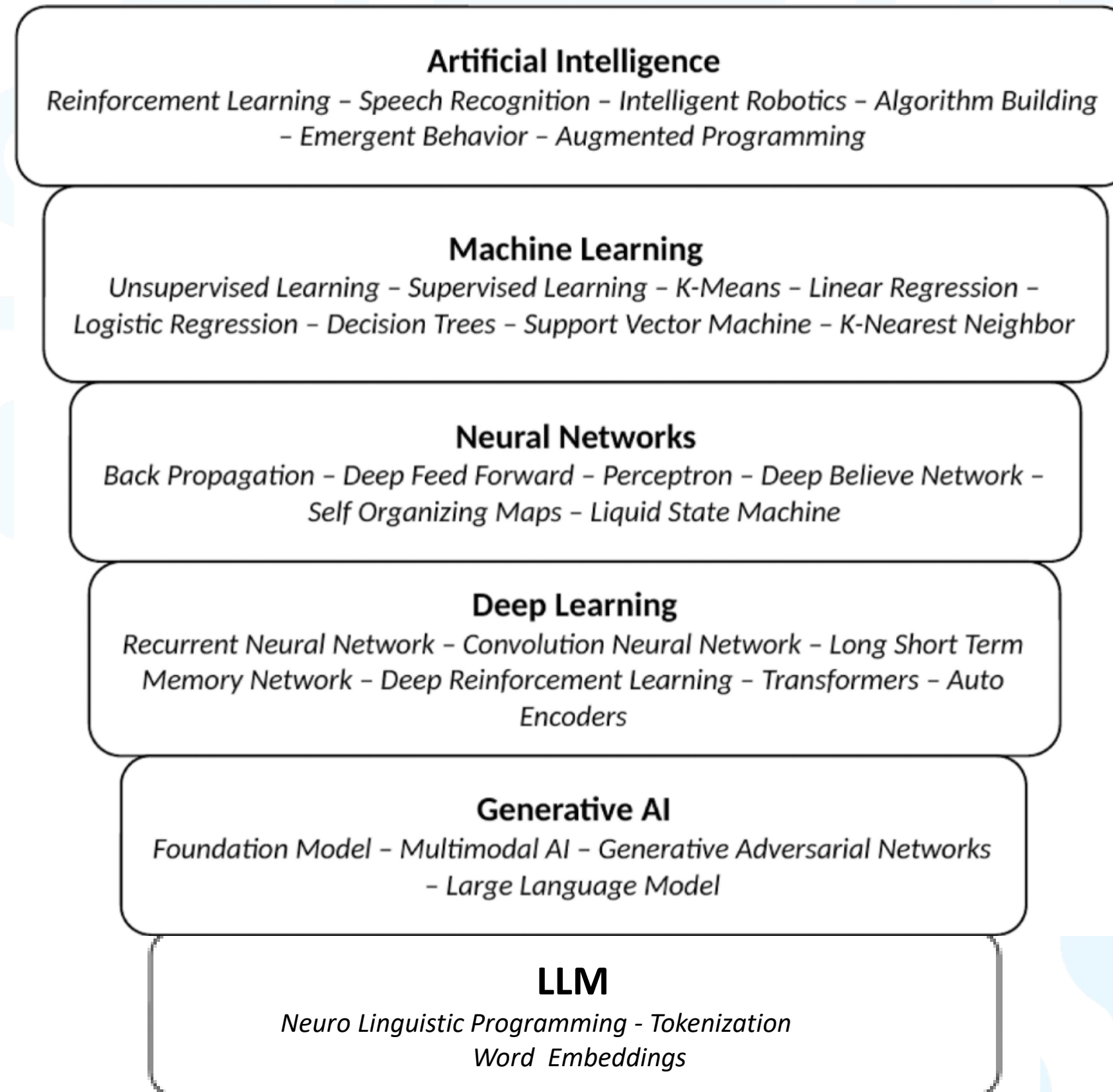
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# Agentic AI



Courtesy of Purdue University





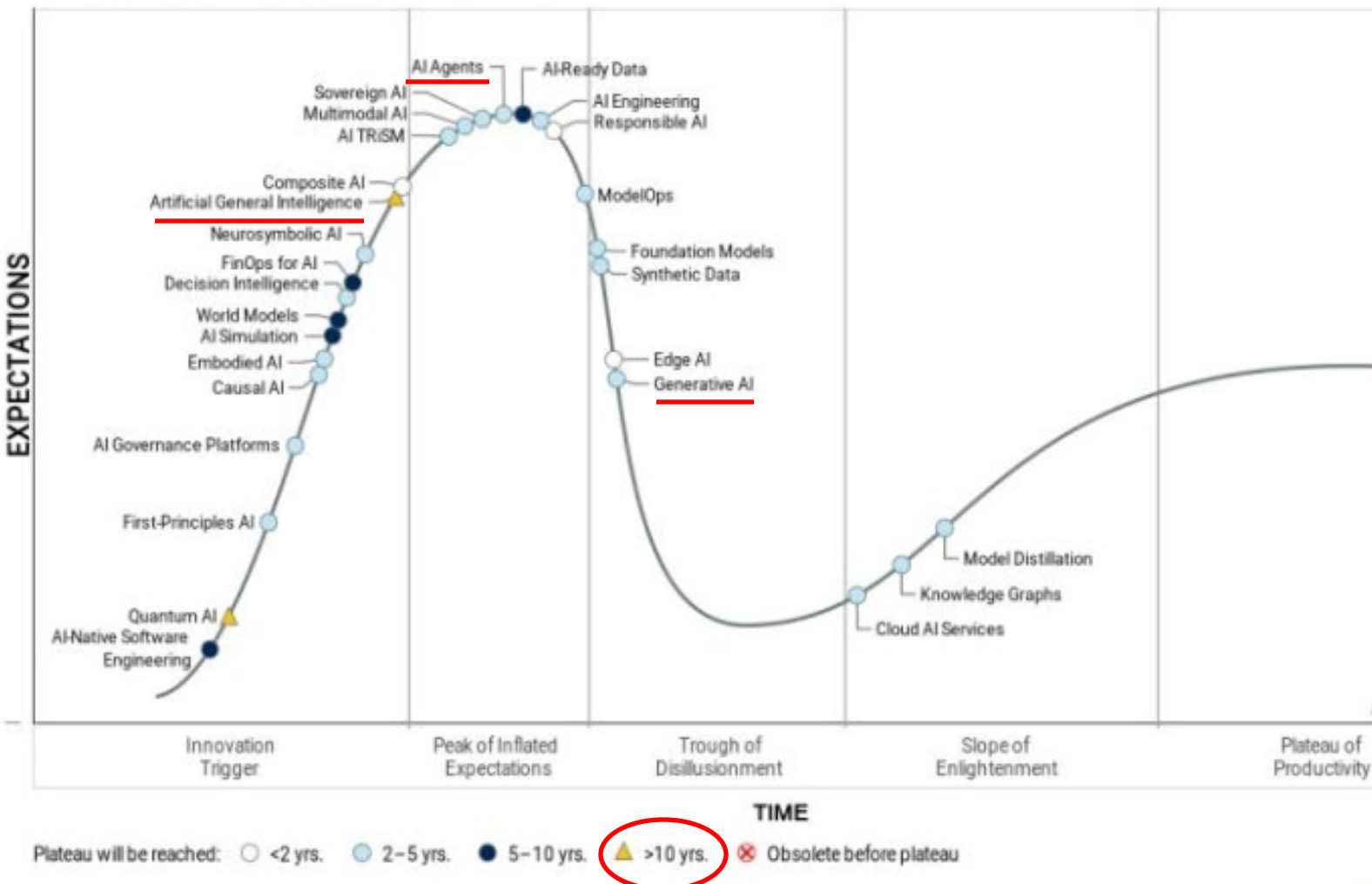
# Challenges of AI

- Hallucinations
- Misinformation
- Anthropomorphism
- Cognitive Offloading
- Data Privacy
- Copyright
- Discrimination
- Fraud
- Environmental Impact

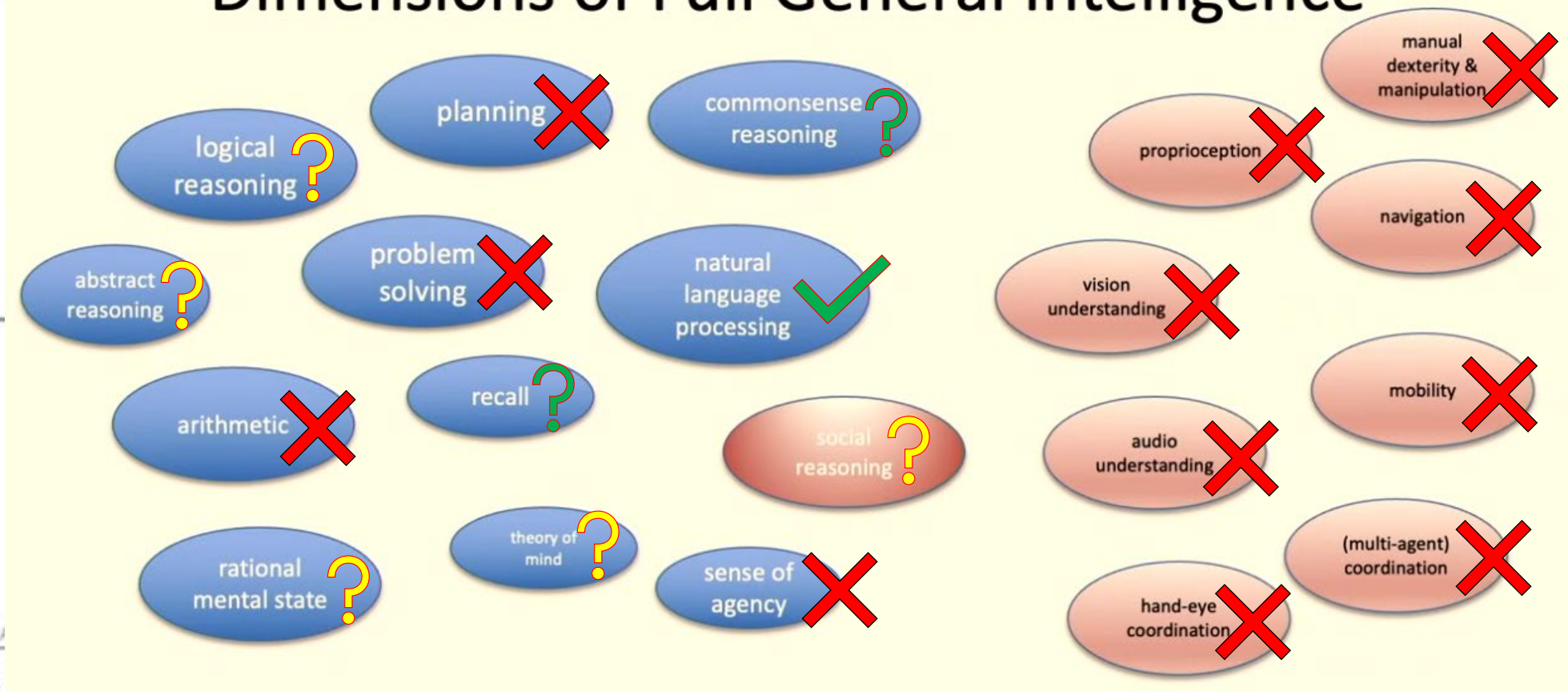


# The Future of AI

Hype Cycle for Artificial Intelligence, 2025



## Dimensions of Full General Intelligence



Courtesy of Prof. Michael Wooldridge

Gartner

## Quiz

1. What is Deep Learning?
2. What is a Word Embedding?
3. How does AI differ from human intelligence?
4. How do you use AI?



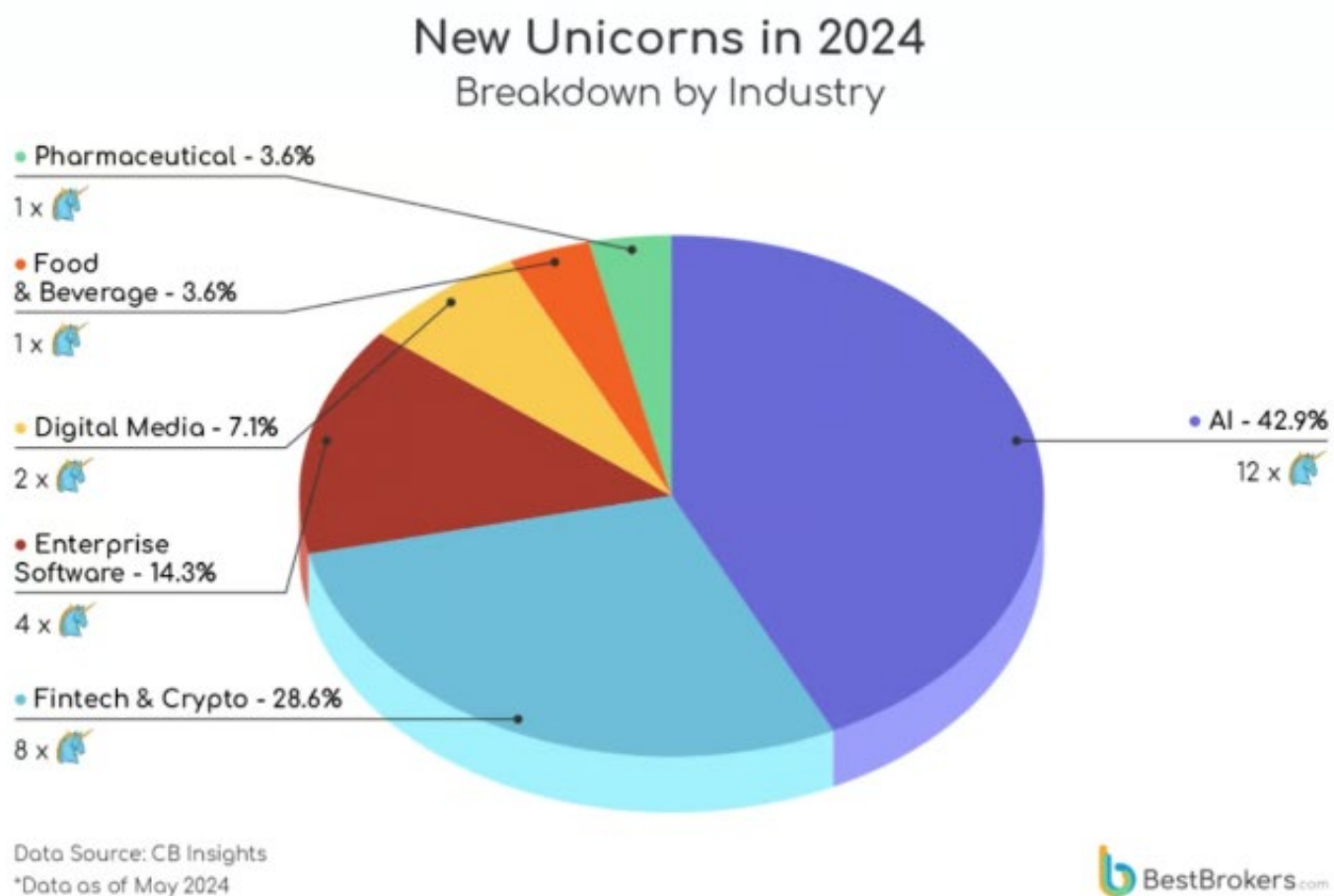


# BACKUP SLIDES



# Growth and Adoption of Generative AI

## Artificial Intelligence Taking Over the Unicorn herd in 2024



## HOW LONG IT TOOK TOP APPS TO HIT 100M MONTHLY USERS

ChatGPT is estimated to have hit 100M users in January, 2 months after it's launch. Here's how long it took other top apps to reach that:

APP	MONTHS TO REACH 100M GLOBAL MAUS
CHATGPT	2
TIKTOK	9
INSTAGRAM	30
PINTEREST	41
SPOTIFY	55
TELEGRAM	61
UBER	70
GOOGLE TRANSLATE	78

SOURCE: UBS

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