



## Applied AI & ML Industry Projects Lab (AAM-IPL)

Transforming Learning into Industry Solutions

### Inaugural Lecture

# Introduction and State of the Art in AI & ML

V Semester - ML and AIUP, Aug-Oct 2024

Session Date and Time: 8<sup>th</sup> Sept 2024, 10:30 AM IST – 12:00 Noon IST

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# Lecture Outline



- What is AI & ML
- The Universe of AI, ML, DS
- AI Capabilities, Types, Sub-fields, Core Areas, Layer Hierarchy
- Real-world AI Applications
- How Do Machines Learn?
- AI Key Concepts
- AI Timeline and 21<sup>st</sup> Century Growth Enablers
- GPTs and LLMs Key Concepts
- State of the Art in AI & ML
- Dangers, Challenges and Limitations of AI
- Future Direction of AI
- AAM-IPL Overview
- Q&A
- Call to Action

# What is AI?



This image is created (2 iterations) by Google Gemini's Imagen 3 for this presentation.  
The prompt provided is "A central image of a brain merging with a digital network (to represent the fusion of human intelligence and artificial intelligence)."

## What makes machines *think*?

- Can machines understand language like we do?
- How do algorithms recognize faces and voices?
- What powers the creativity of AI-generated art?
- Is AI just a tool, or could it ever be more?

## Artificial Intelligence is the New Electricity!

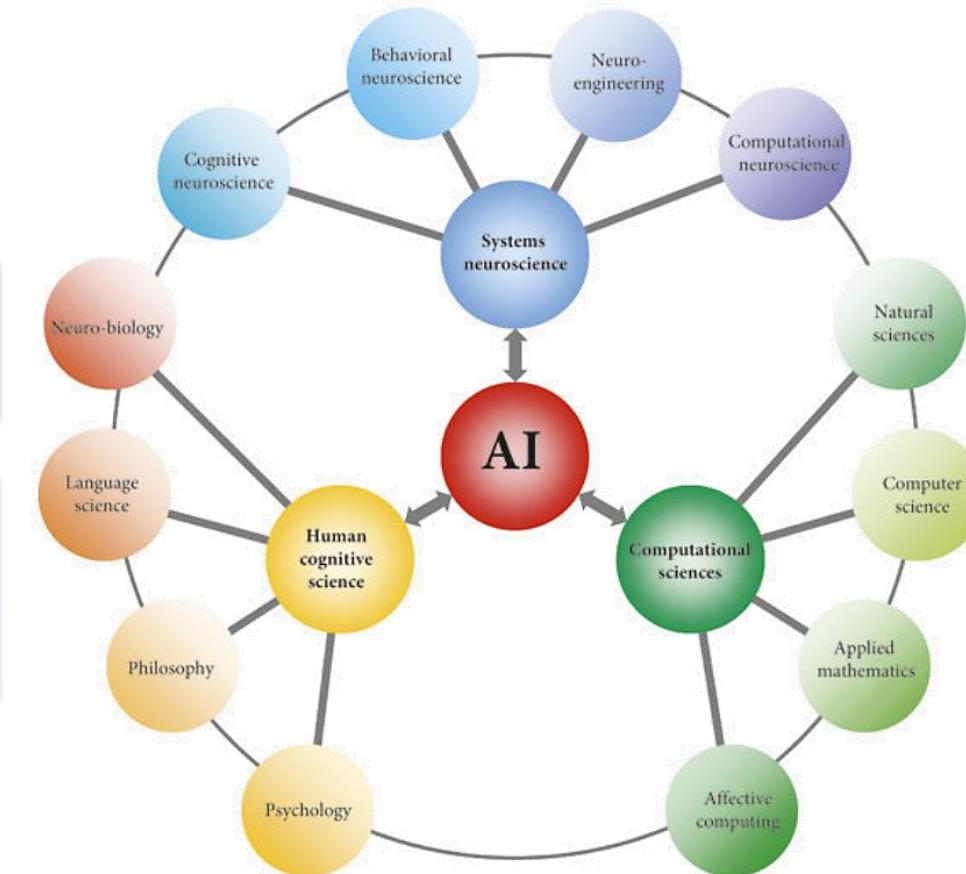
- Andrew Ng, AI Pioneer, Adjunct Professor at Stanford University, Co-founder and Chairman of Coursera

# AI – Multi-disciplinary Science

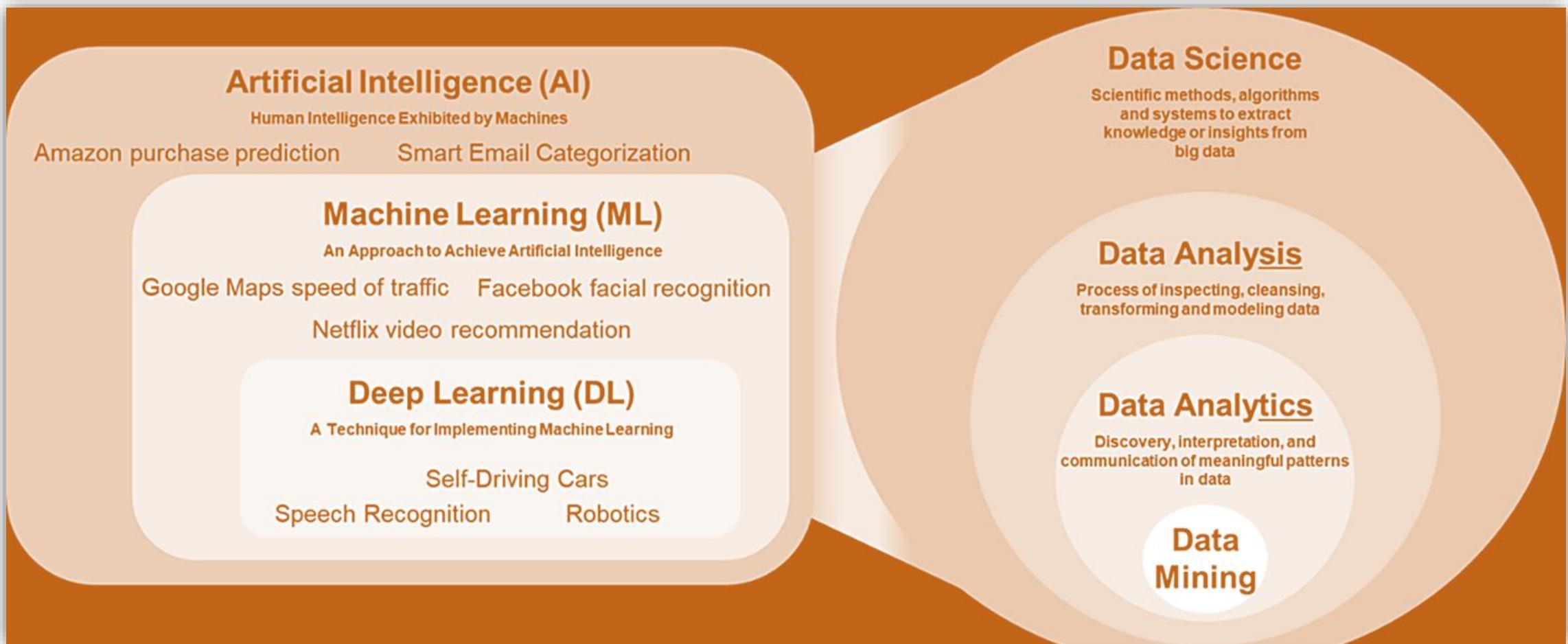
Branch of computer science and engineering and intersection of many other sciences

- Mimic human cognitive functions - such as learning, problem-solving, and decision-making

- Learn, reason, solve real-life or business problems, perceive things, understand natural languages like English, make decisions etc.



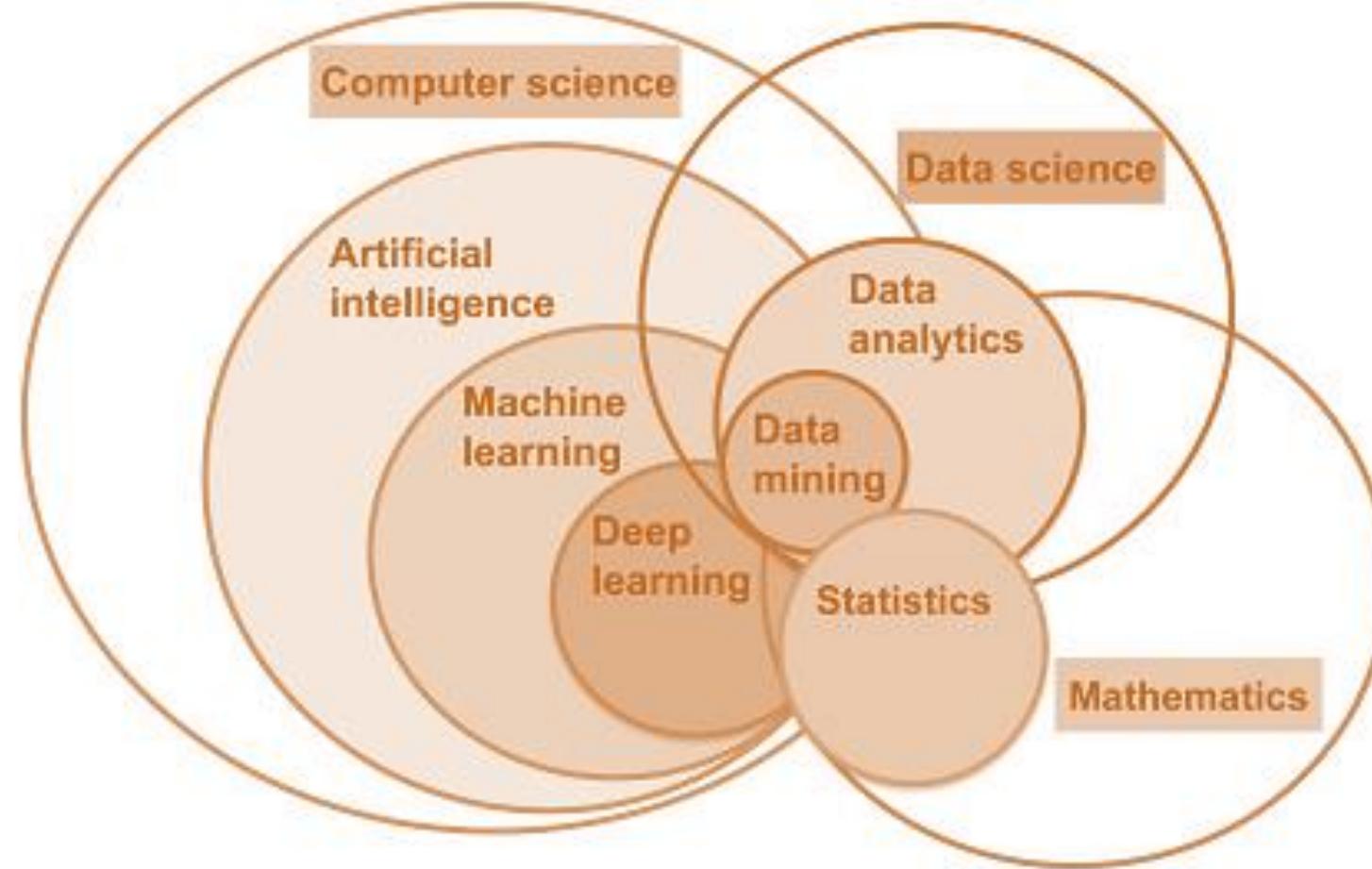
# The Universe of Data Science and Artificial Intelligence



**Big Data**



# The AI Intersection of Sciences



# AI Capabilities and Types

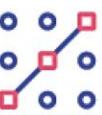
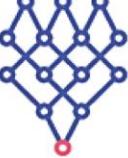
## AI Capabilities

**Artificial Narrow Intelligence (ANI):** Performing narrowly defined sets of tasks

**Artificial General Intelligence (AGI):** Simulating thought processes of the human mind

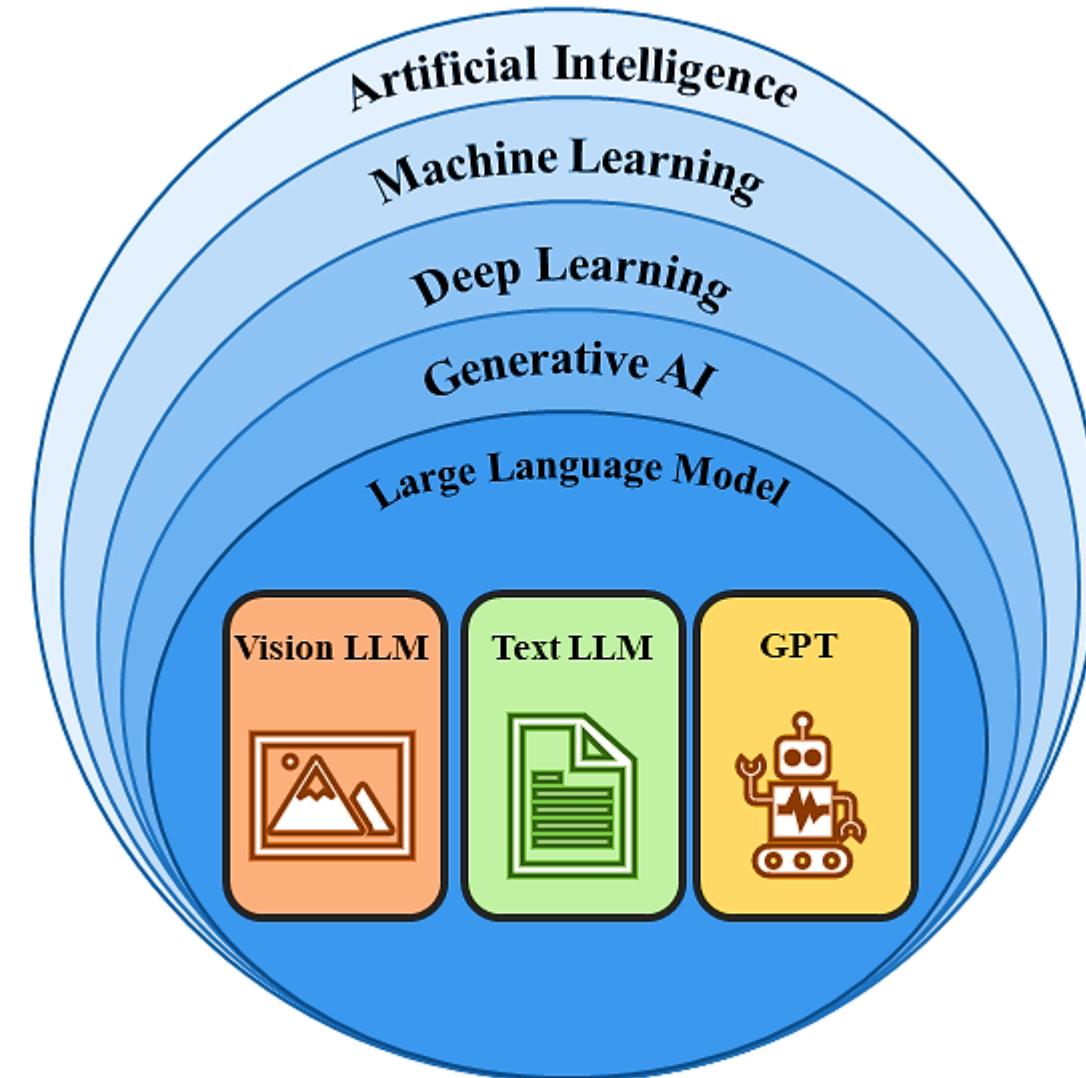
**Artificial Super Intelligence (ASI):** Performing beyond human capability

## AI Systems<sup>1</sup>

Reactive AI	Limited memory	Theory of mind	Self-aware
<ul style="list-style-type: none"><li>○ Good for simple classification and pattern recognition tasks</li><li>○ Great for scenarios where all parameters are known; can beat humans because it can make calculations much faster</li><li>○ Incapable of dealing with scenarios including imperfect information or requiring historical understanding</li></ul> 	<ul style="list-style-type: none"><li>○ Can handle complex classification tasks</li><li>○ Able to use historical data to make predictions</li><li>○ Capable of complex tasks such as self-driving cars, but still vulnerable to outliers or adversarial examples</li><li>○ This is the current state of AI, and some say we have hit a wall</li></ul> 	<ul style="list-style-type: none"><li>○ Able to understand human motives and reasoning; can deliver personal experience to everyone based on their motives and needs</li><li>○ Able to learn with fewer examples because it understands motive and intent</li><li>○ Considered the next milestone for AI's evolution</li></ul> 	<ul style="list-style-type: none"><li>○ Human-level intelligence that can bypass our intelligence, too</li><li>○ Considered a long-shot goal</li></ul> 

1: Prof. Arend Hintze's (Dalarna University, Sweden) Hypothesis

# Sub-fields of AI



# Core Areas - Artificial General Intelligence

AI/GAI

AL, DL, ANN

Computer Vision

Robotics, Intelligent Agents

NLP

TTS and STT

Expert Systems, Knowledge Representation and Reasoning

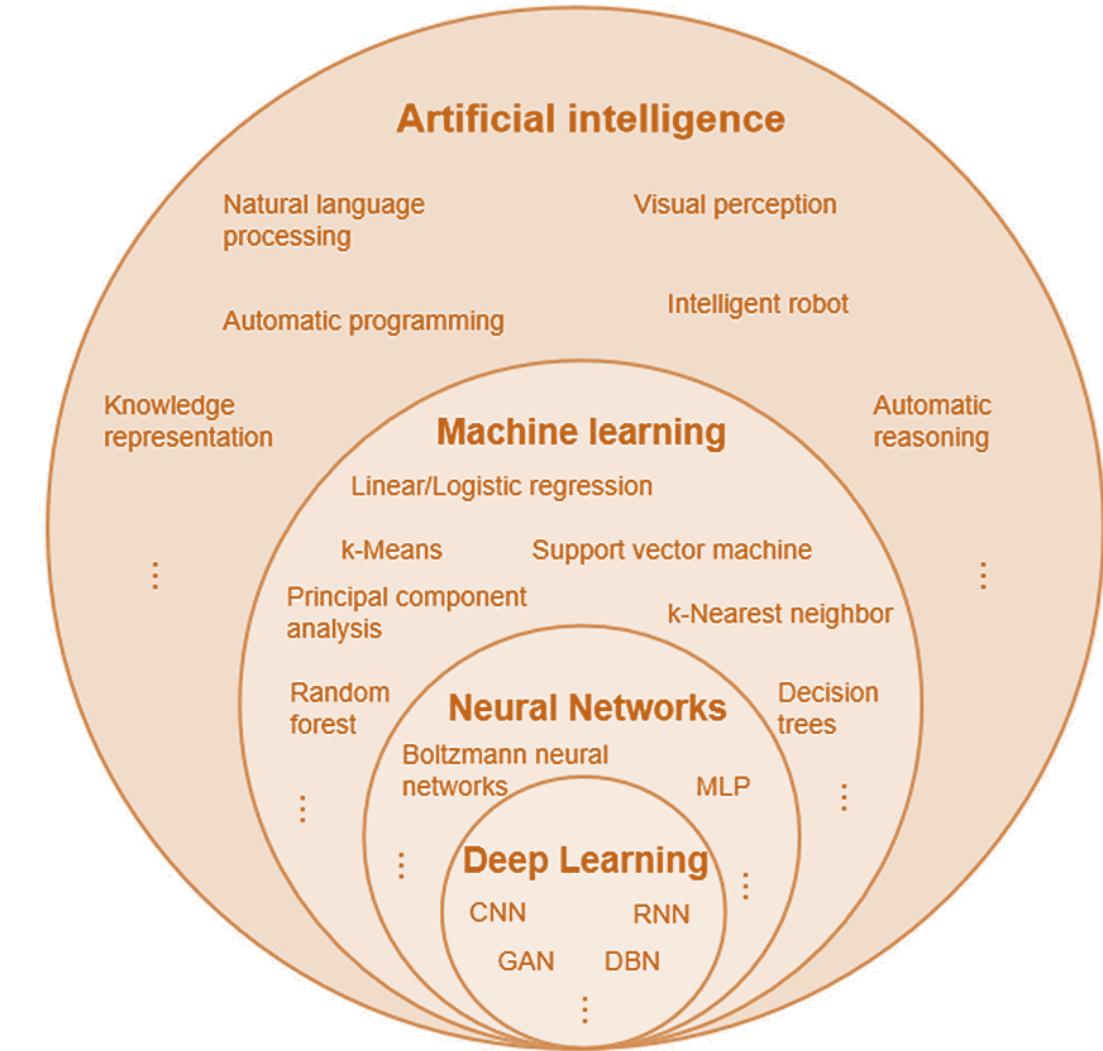
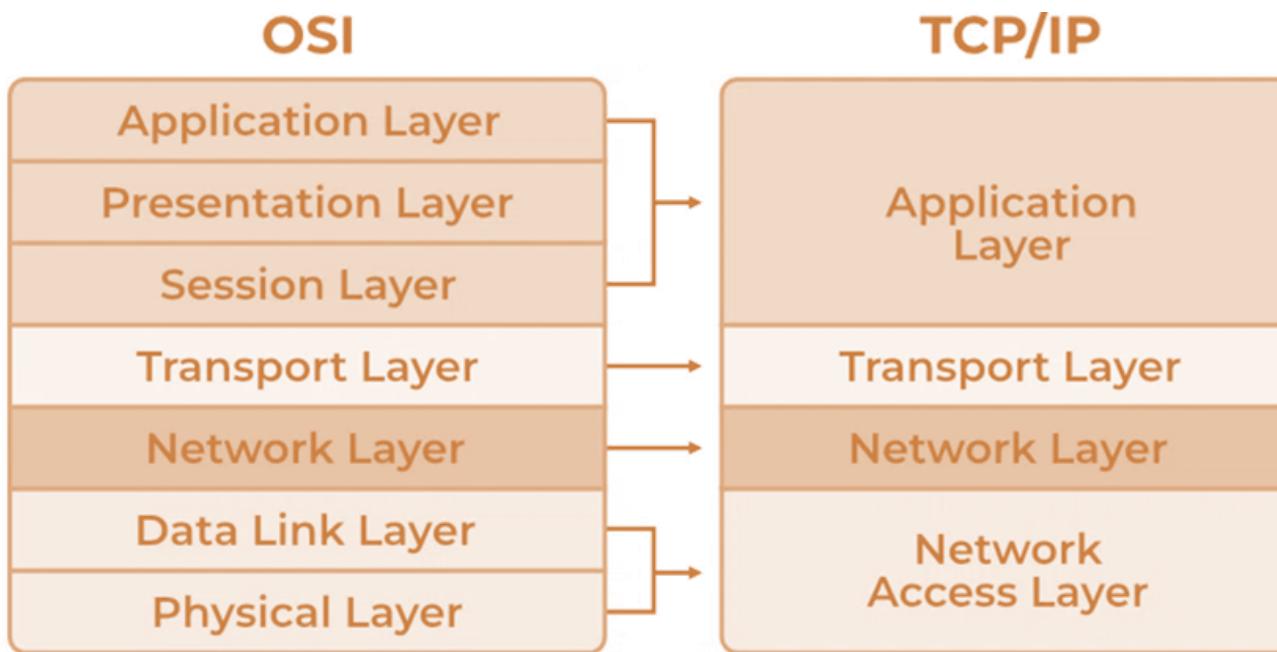
Autonomous Vehicles

Augmented and Virtual Reality

Social Networks and Graph Analysis



# AI Layer Hierarchy



# Real-world Applications of AI

## Healthcare

- ML algorithms are facilitating early disease detection, accurate diagnosis, personalized medicine etc.

## CRM

- NLP powered virtual assistants, chatbots providing 24\*7 uninterrupted human-free service
- ML algorithms to identify buyers' behaviour, preferences to provide **hyper-personalized customer experience**
- Algorithms can automatically generate customized product recommendations, promotions and content for customers and prospects

## Finance

- ML algorithms are employed for fraud detection, algo trading, credit scoring and risk assessment
- AI-powered financial planning and wealth management platforms, robo-advisors leverage advanced algos to analyze markets trends, assess client risk tolerance, provide personalized investment recommendations

## Manufacturing

- Advanced algos to detect defects in products to ensure quality, predictive maintenance, supply chain optimization, robotics

## Transportation

- Self-driving cars, route optimization, public transport passenger demand prediction, schedules optimization etc.



# Real-world Applications of AI

## Agriculture

- Precision farming with AI farmers - data-driven decisions for irrigation optimization, improved fertilization and reduced waste.
- AI to monitor crops for infestation and appropriate pesticides, identify predict yields etc.
- Self-driven tractors for farming, drone-driven pesticide sprinkling and spraying etc.

## Retail

- Personalized shopping experience, personalized product recommendations, just-in-time inventory,

## Education

- AI-driven language translation, real-time transcription and sub-titles
- Automatic grading of answers/solution

## Energy

- Smart grid management, demand forecasting, predict usage patterns for efficient handling of power generation/sourcing etc.

## Human Resources

- Automatic resume screening, advanced algos for creating, assessing job posts etc.

## Business Intelligence

- Insights and recommendations generated by AI models can help take data-driven decision-making for businesses efficiency and growth



# Real-world Applications of AI

## Government

- AI is being used to improve public safety, detect crime, and provide citizen services

## Security

- AI algos for facial recognition, surveillance, and threat detection

## Entertainment

- AI-generated content for immersive experience – VR, AR
- AI-powered recommendation and content curation for engaging and personalized entertainment experience

## Law and Legal

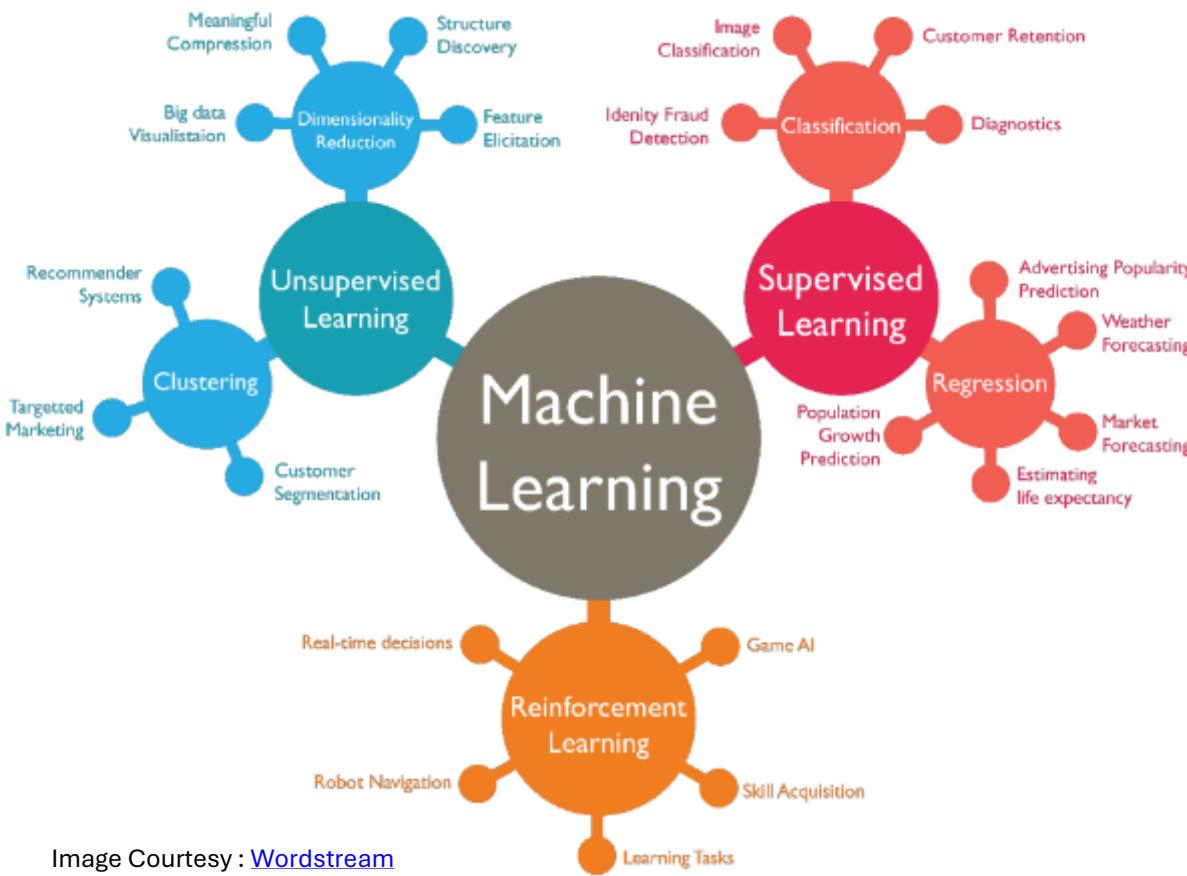
- Identifying potential issues in contracts, legal documents, and expediting legal research by extracting relevant information using AI-driven tools

## Space

- AI is employed for space craft navigation, satellite imaging, mission planning, new astronomical phenomena identification etc.



# How Do Machines Learn?

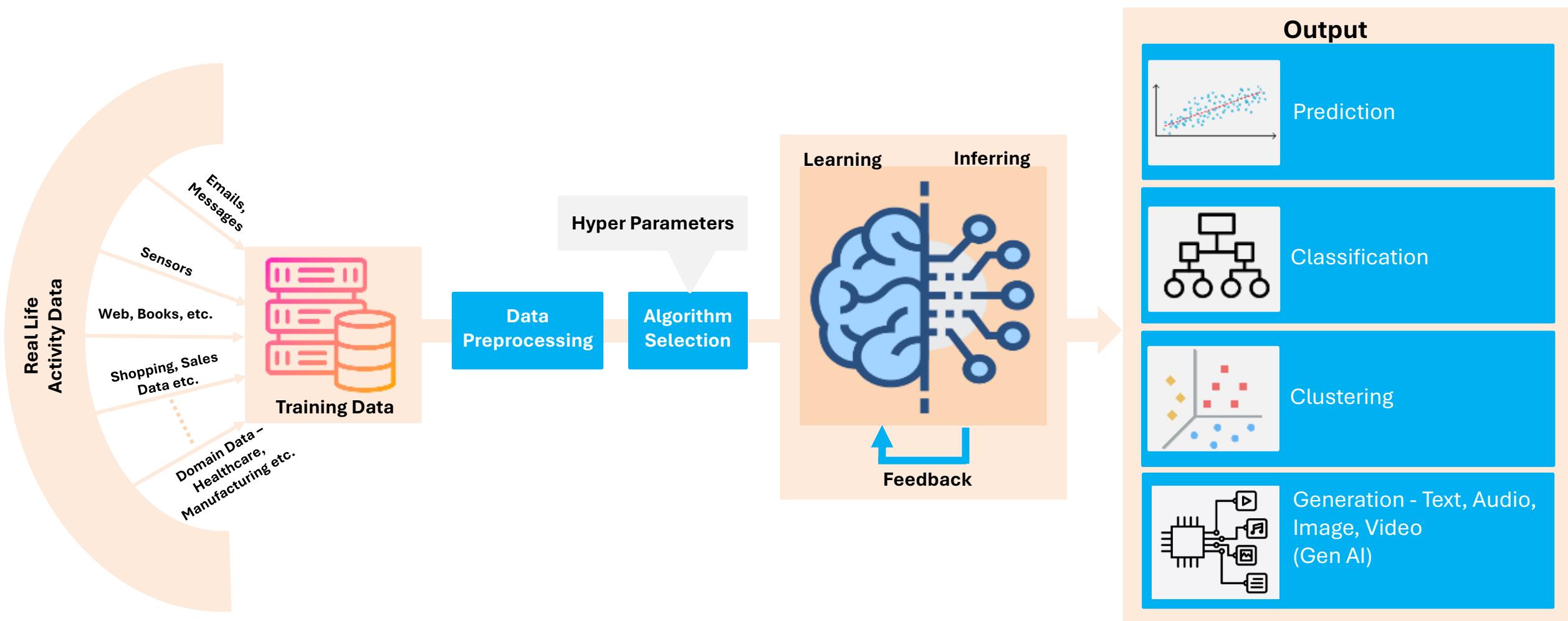


- Can a computer learn without explicit instructions?
- How do AI models make sense of the vast amounts of data they are fed?
- What makes a machine get better over time?

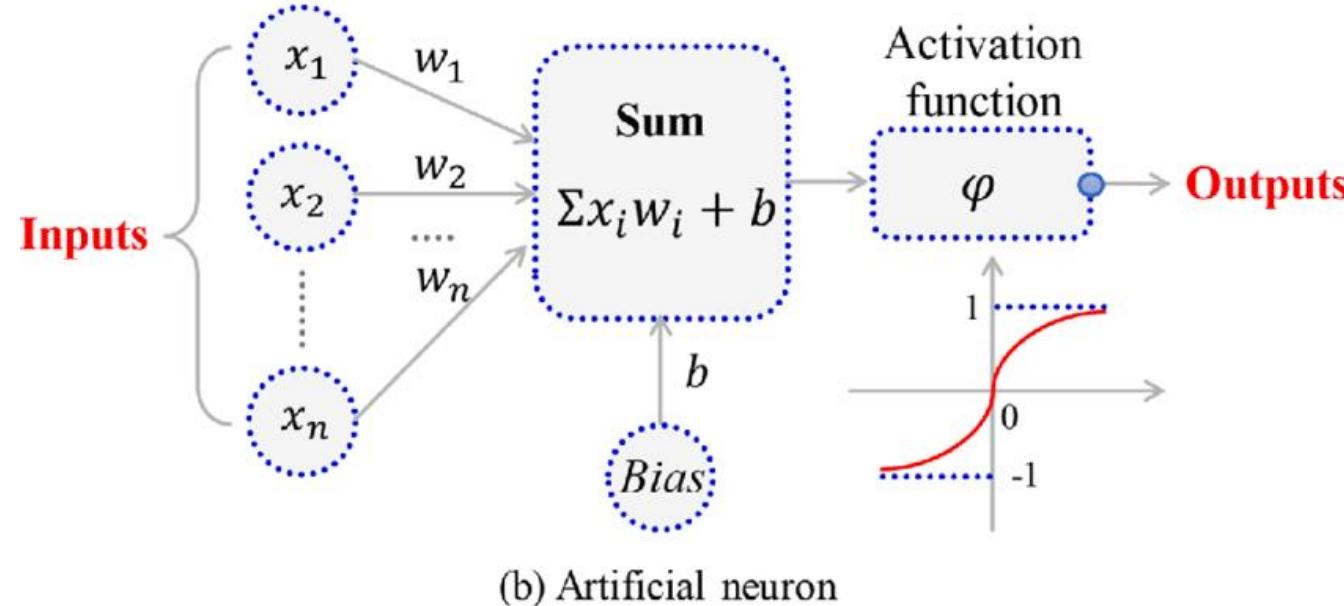
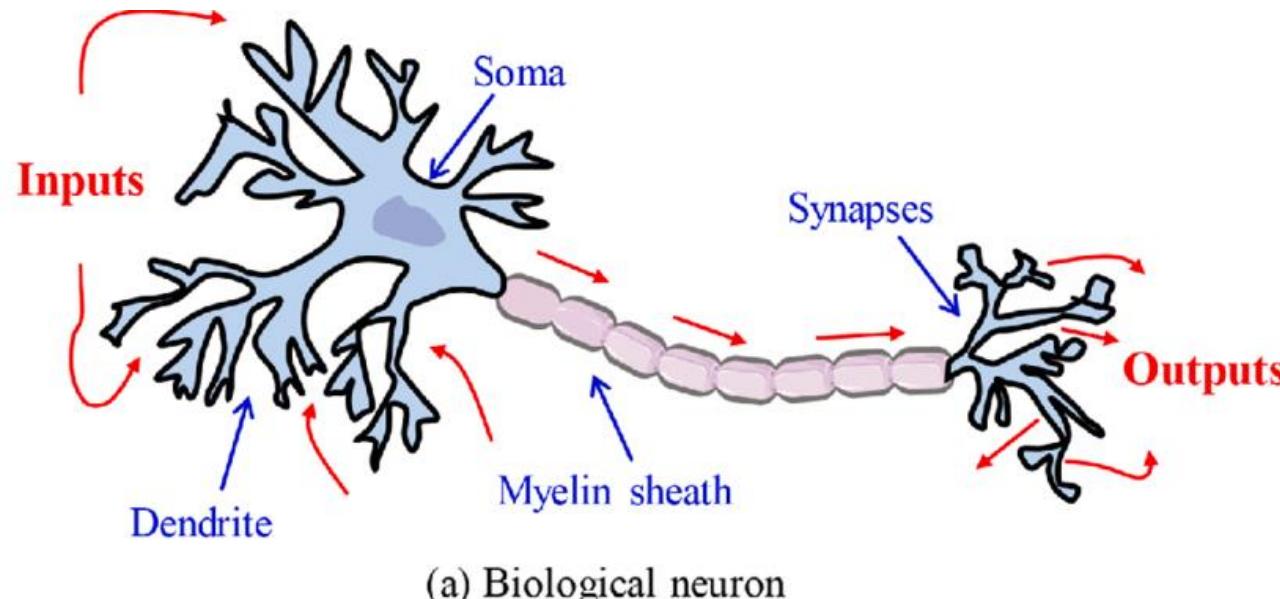
**We don't need to teach a machine to think like us; we need to teach it to learn from us**

- Yann LeCun, Deep Learning Pioneer, Chief AI Scientist at Meta and Professor at New York University

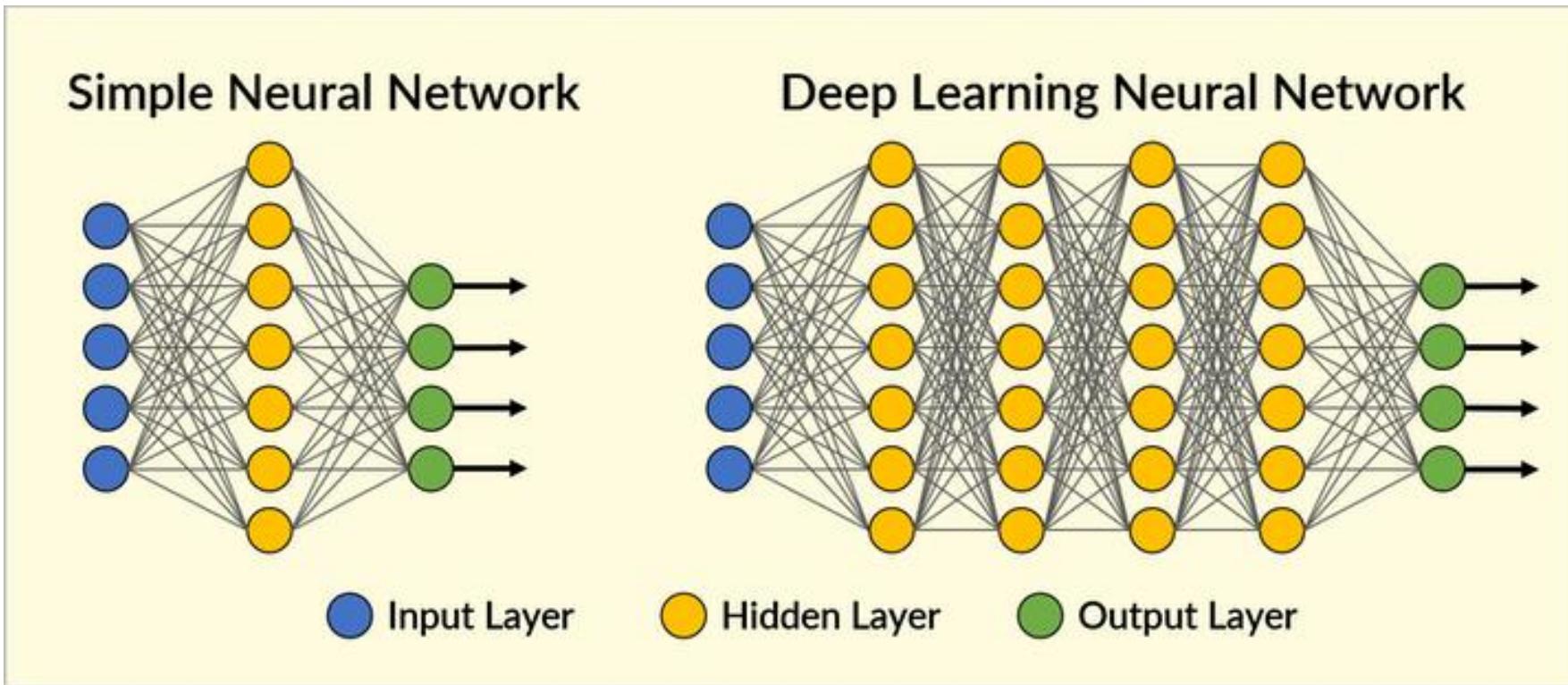
# How Do Machines Learn? (Contd..)



# AI Key Concepts - Biological and Mathematical Neurons

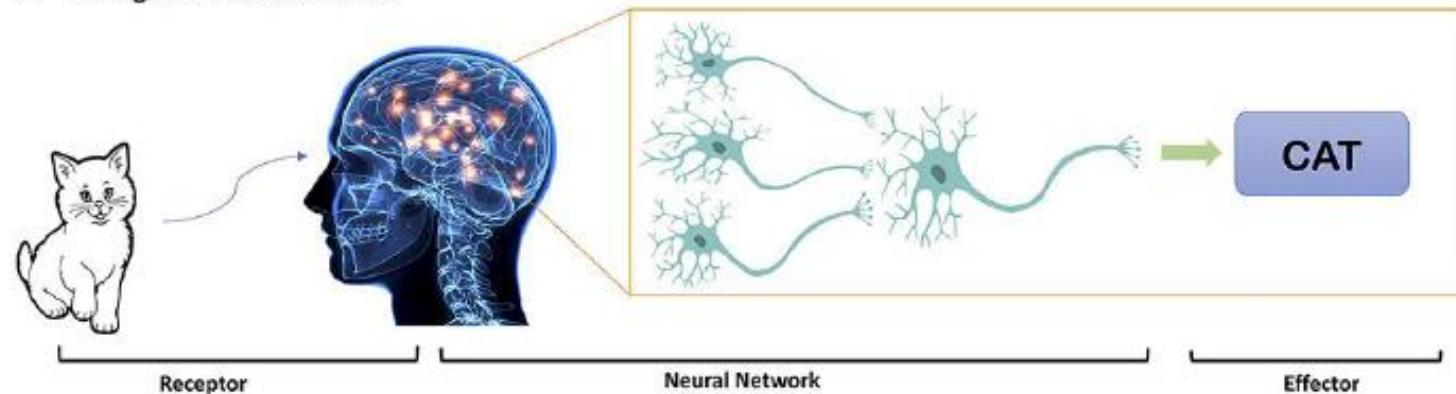


# AI Key Concepts - Deep Learning

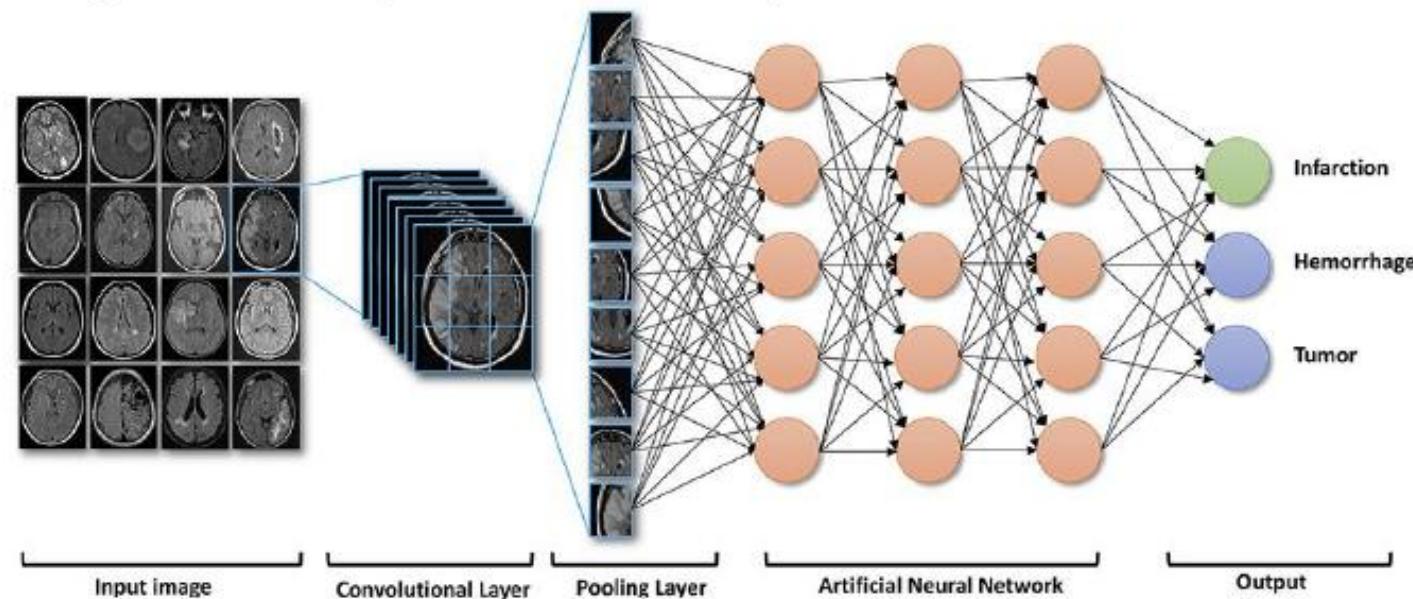


# AI Key Concepts - Deep Learning

A Biological Neural Network

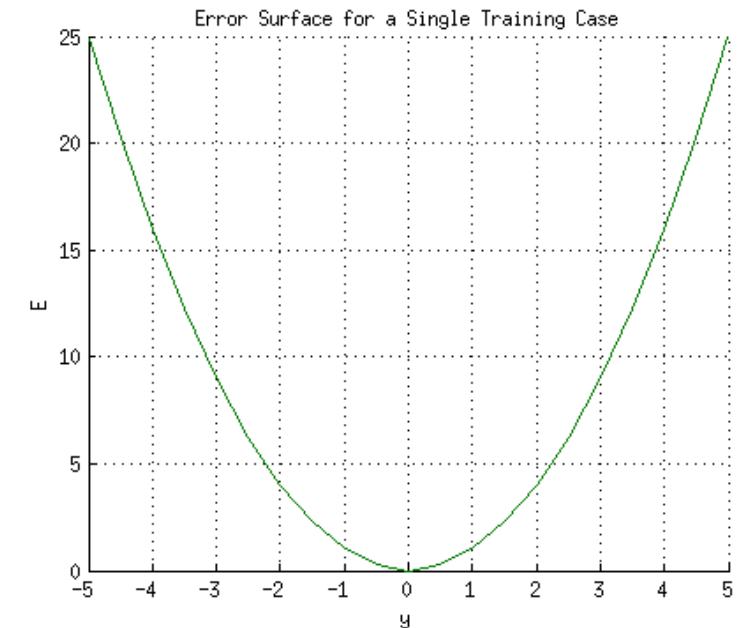


B Computer Neural Network(Convolutional Neural Network)

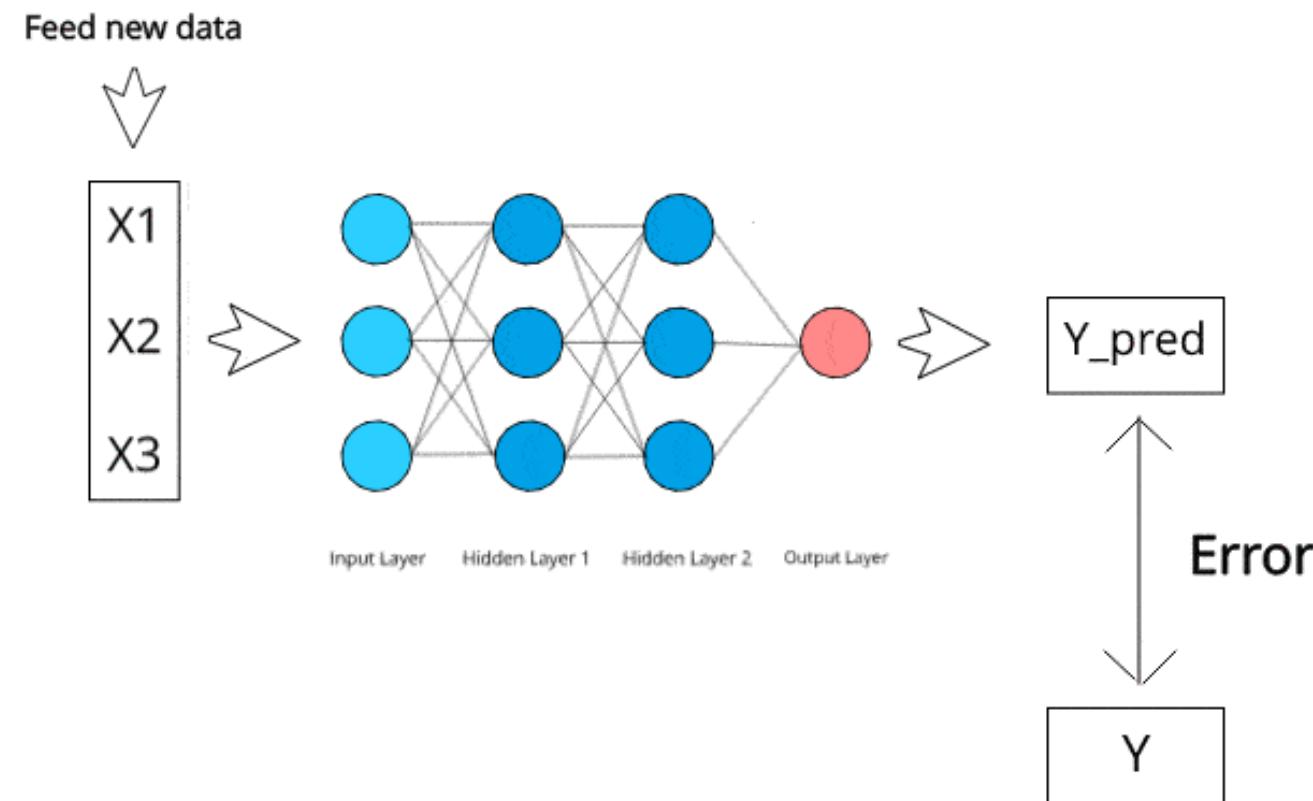


# AI Key Concepts - Back-propagating Error Correction aka Backpropagation

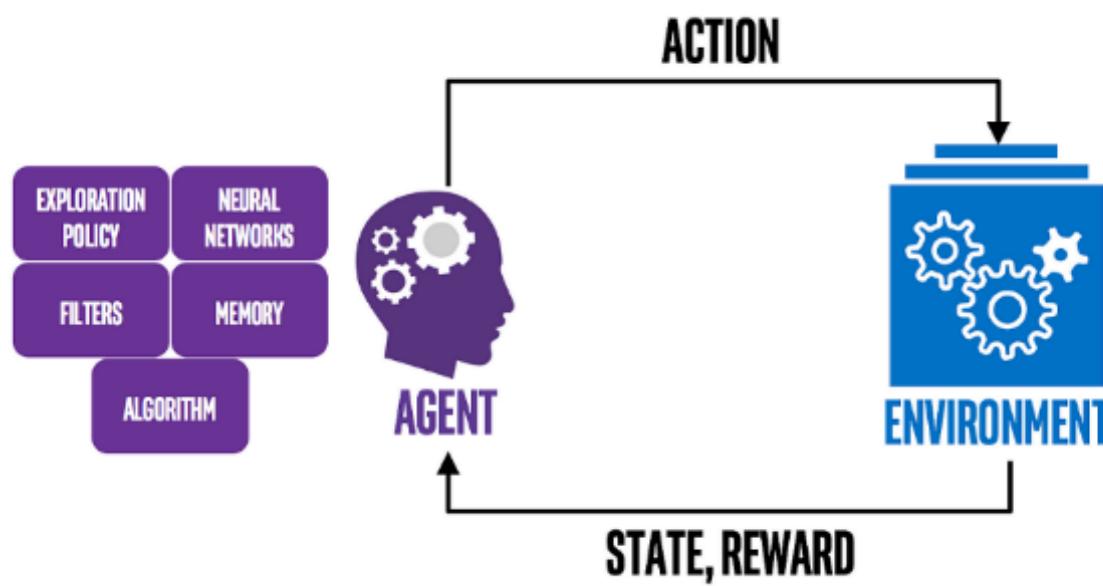
- Goal of any supervised learning algorithm (ML or DL) is to find a function that best maps or fits set of inputs ( $x_1, x_2$ ) to their outputs ( $t$ )
- Initially, with random weights, the predicted output  $y$  which is likely to differ from  $t$  –  $y = w_1 * x_1 + w_2 * x_2$
- Consider one-shot training case with tuple (1,1,0)
- The loss or cost function  $\mathcal{L}(t, y) = E = (t - y)^2$  - Squared Error
- The minimum of “**E vs y plot**” corresponds to output  $y$  which minimizes the error  $E$  (to 0)
- **Finding the mapping or learning function** boils down to an **optimisation problem** of **finding a function that will produce minimal error**
  - $f(\cdot) = \min_{s.t.y=(w_1*x_1+w_2*x_2)} \mathcal{L}(t, y)$
- Backpropagation is a gradient estimation method during training neural networks to compute the network parameter updates
- It is a chain rule that computes the gradient of the loss function with respect to weights of the network for single input-output - one layer at a time iterating backward from the last layer
- Backpropagation refers only to an algorithm for efficiently computing the gradient, not how the gradient is used



# AI Key Concepts - Back-propagating Error Correction aka Backpropagation (Contd...)

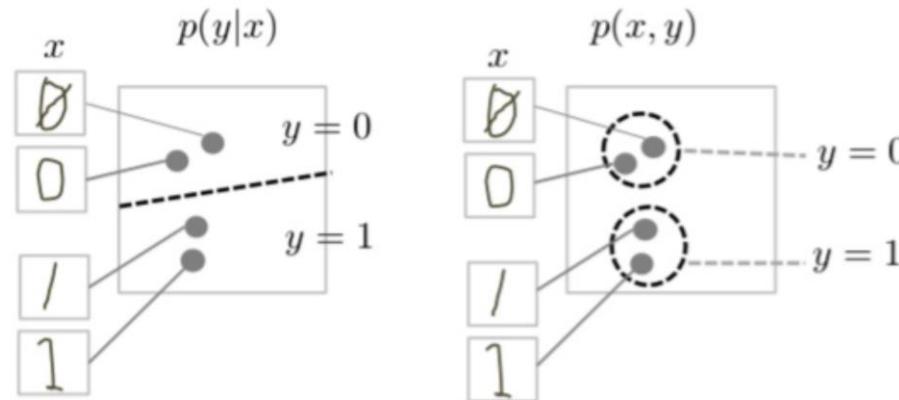


# AI Key Concepts - Reinforcement Learning



# AI Key Concepts - Generative Vs Discriminative Models

- Discriminative Model
- Generative Model



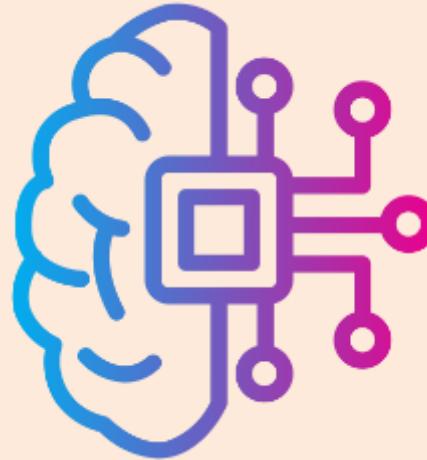
Aspect	Discriminative Models	Generative Models
Objective	Model the conditional probability ( $P(Y   X)$ )	Model the joint probability $P(X,Y)$
Focus	Learn the boundary that separates different classes	Understand the underlying distribution of data
Examples	Logistic Regression, SVM, Neural Networks, Random Forests	Naive Bayes, Gaussian Mixture Models, HMM, VAEs
Advantages	Better classification performance, directly optimizes decision boundaries	Can generate new data, handles missing data well
Disadvantages	Cannot generate new data, requires a lot of labeled data	Requires assumptions about data distribution, less accurate for classification tasks
Usage	Classification, regression	Data generation, density estimation

## Artificial Intelligence



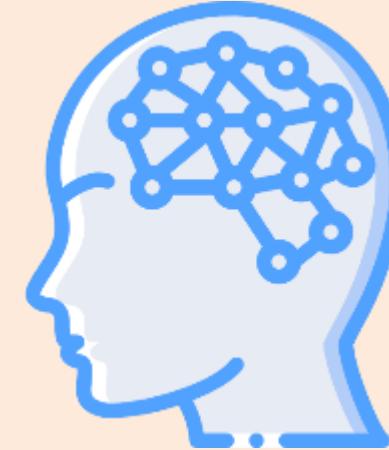
Engineering intelligent machines and programs

## Machine Learning



Making computers learn without being explicitly programmed

## Deep Learning



Learning based on Deep Neural Networks and vast amounts of data

## GPTs and LLMs



Building large foundational pre-trained language models to recognize, translate, predict, or generate text or other content

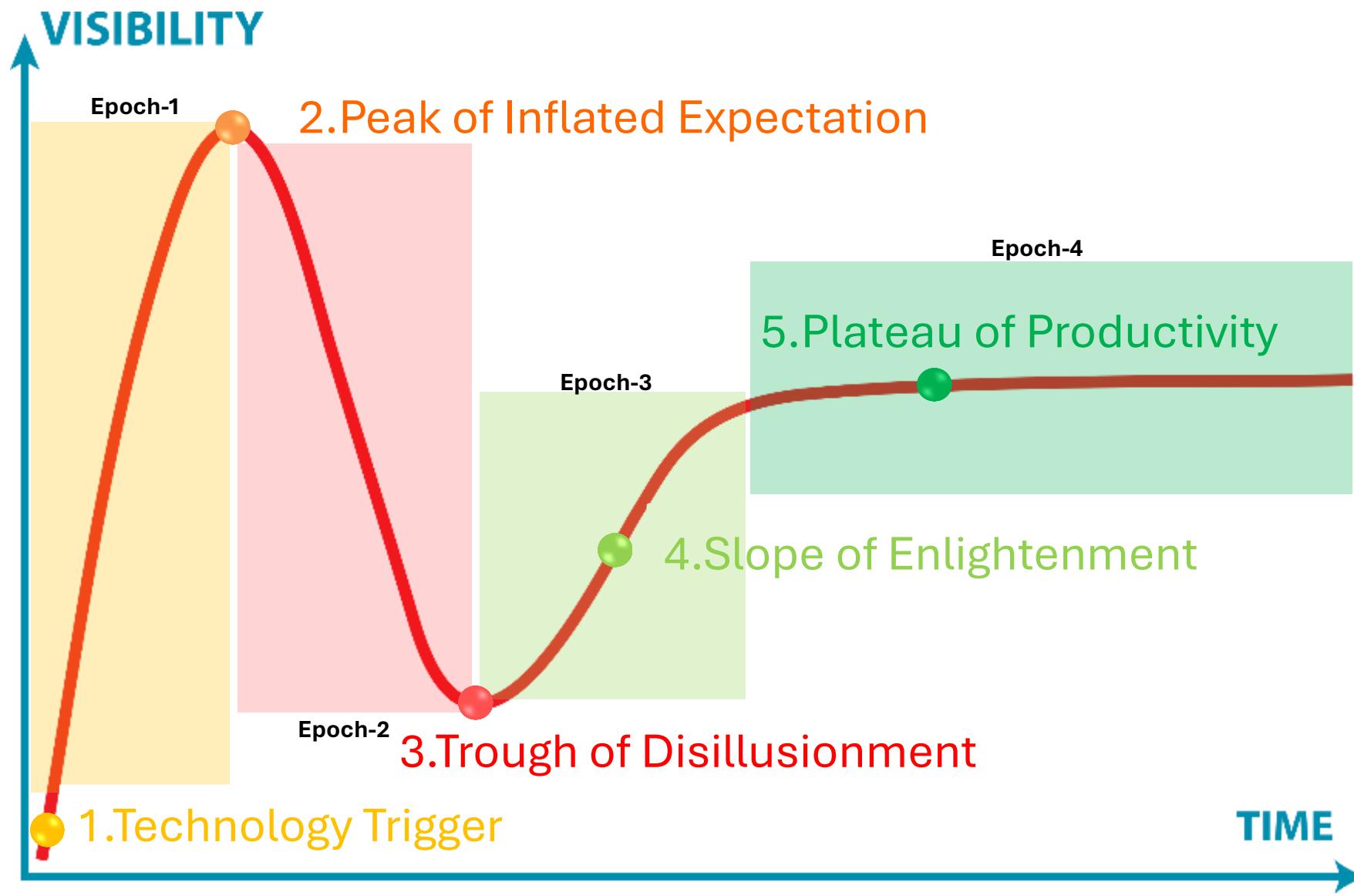
1950 – 1970

1980 - 2006

2010 - 2017

2018 - 2024

# New Technology Journey in Business



Created and popularized by Gartner as **Gartner Hype Cycle**



# Key Enablers for AI's Growth in 21<sup>st</sup> Century

Advances in computing power – GPU, TPU, AI Chips

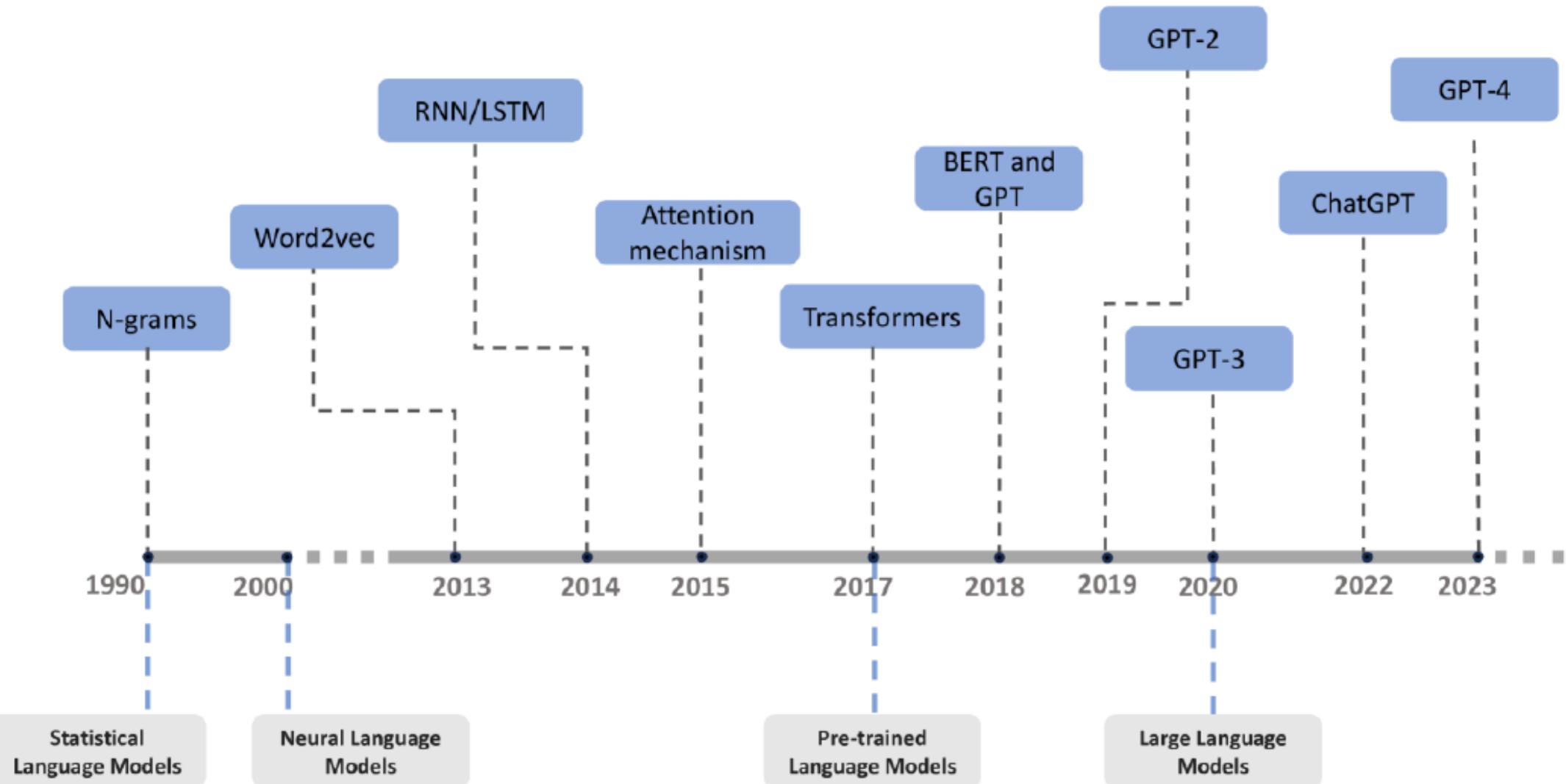
Cloud, User Generated Content, Vast Amounts of Web Data and Digital Assets

Continuous Improvements in AI

Advances in NLP – Transformer, GPTs, LLMs



# NLP, GPT and LLMs Timeline



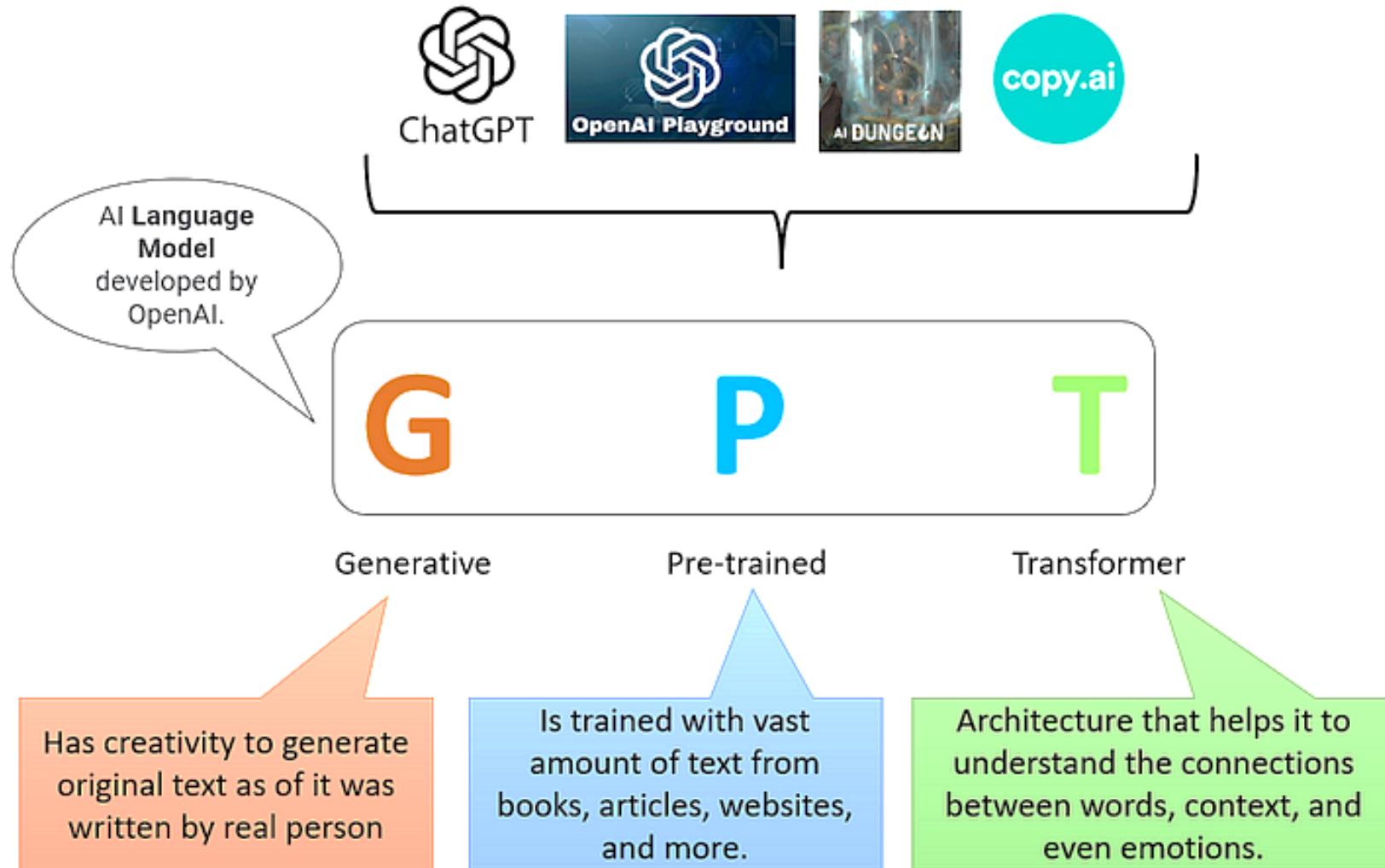
# GPTs and LLMs

## GPTs – Generative Pre-trained Transformers

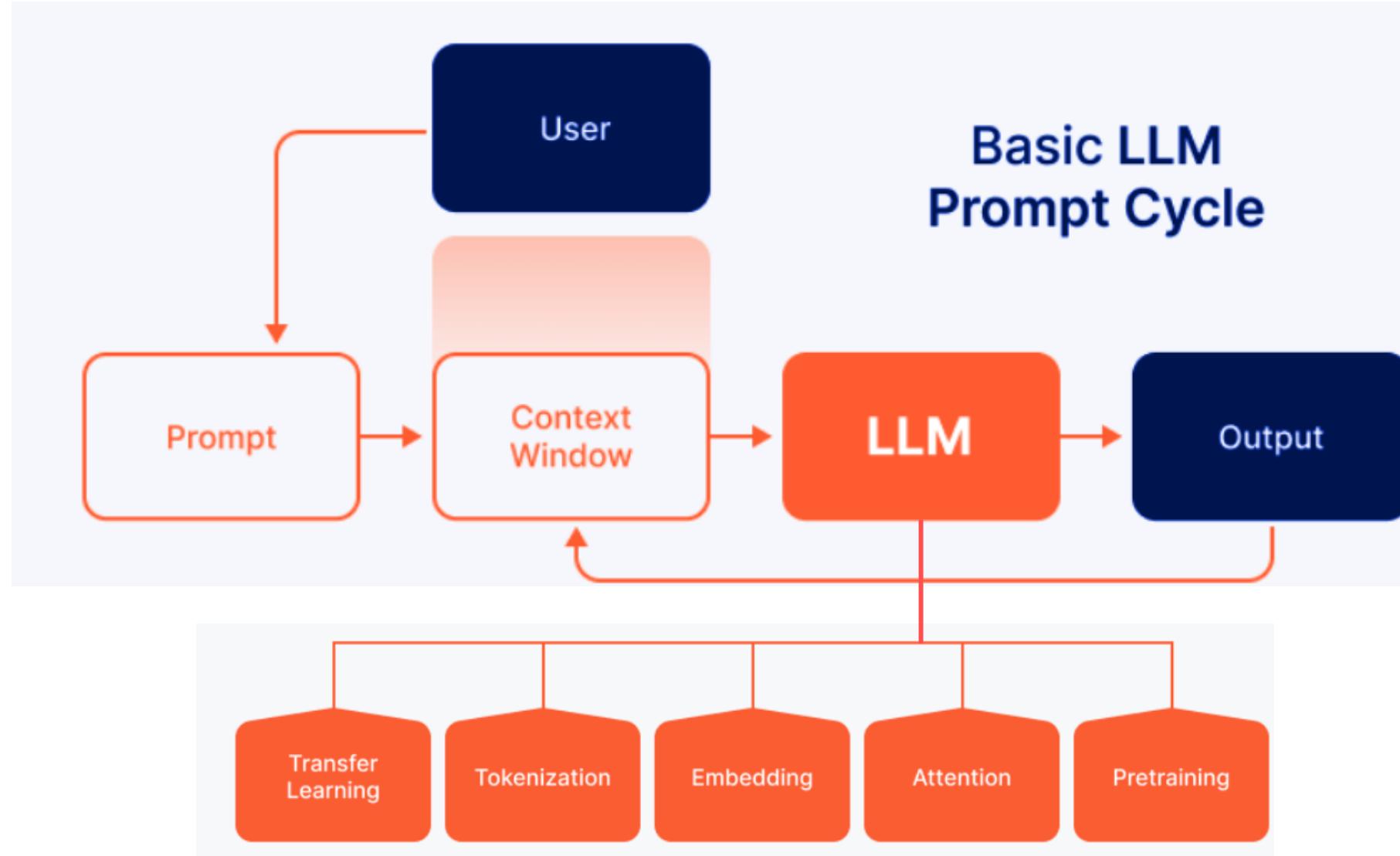
- Specific type of LLM developed by OpenAI is designed specifically for generating coherent and contextually relevant text.
- Example : GPT-1, GPT-2, GPT-3, and GPT-4 of OpenAI
- Generates human-like text given some input or prompt.
- Pre-trained on Peta bytes of text in an unsupervised manner and fine-tuned
- Unidirectional: Traditional GPT models are autoregressive, meaning they generate text one token at a time, considering previous tokens

## LLMs – Large Language Models

- Language model trained on a large corpus of text data designed to understand, generate, and manipulate human language
- Examples: GPT (all versions), BERT, T5, LLaMA, and PaLM
- Have large number of parameters - ranging from millions to hundreds of billions
- **Trained on** vast amounts of text data from the internet, books, articles, etc.
- **Perform variety of NLP tasks** - text generation, translation, summarization, sentiment analysis, question answering, and more.



# LLM Functional Architecture

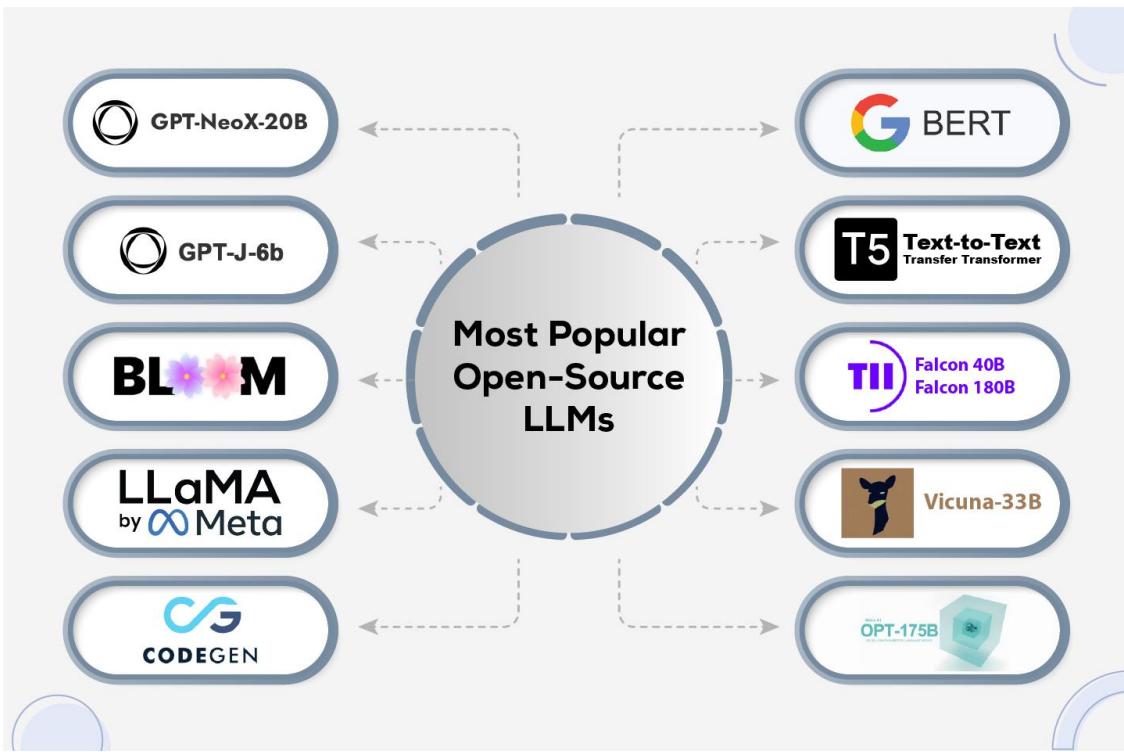


# LLM Compute Requirements

Model	Parameters	Training Data	Training	Inference
BERT	340 million	3.3 million words	$6.73 \times 10^{18}$ FLOPS	680 GFLOPS
GPT-J	6 billion	402 billion tokens	$1.45 \times 10^{22}$ FLOPS	12 TFLOPS
GPT-3	175 billion	300 billion tokens	$3.14 \times 10^{23}$ FLOPS	350 TFLOPS

Computational effort required for training (all training data) and inference (1024 tokens) for a number of popular text models.

# LLMs – Closed and Open Source



Closed Source LLMs				
GPT-3 and GPT-4	Megatron Turing NLG	ERNIE 3.0	Titan	Jurassic-1
Wu Dao 2.0	HyperCLOVA	Gopher	Chinchilla	Galactica
LaMDA	AlexaTM	BloombergG PT	PanGu	Kosmos-1
Sarvam's OpenHathi-Hi-v0.1	Ola's Krutrim LLM	BharatGPT		

# GPT-3 – Training Corpus

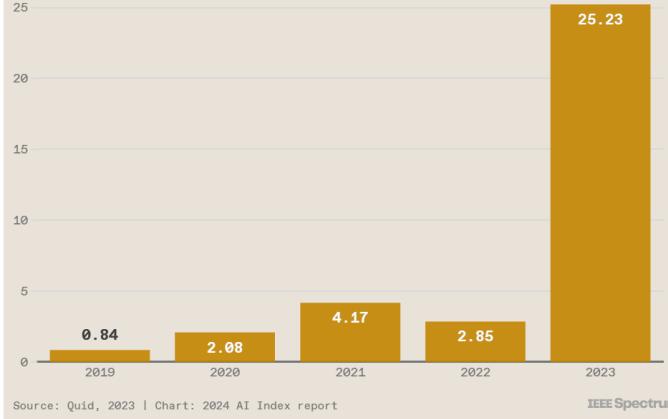


Datasets	Quantity (Tokens)	Weight in Training Mix	Epochs elapsed when training for 300 BN tokens
Common Crawl (filtered)	410 BN	60%	0.44
WebText2	19 BN	22%	2.90
Books1	12 BN	8%	1.90
Books2	55 BN	8%	0.43
Wikipedia	3 BN	3%	3.40

# AI State of the Art - 2024

Private investment in generative AI, 2013–23

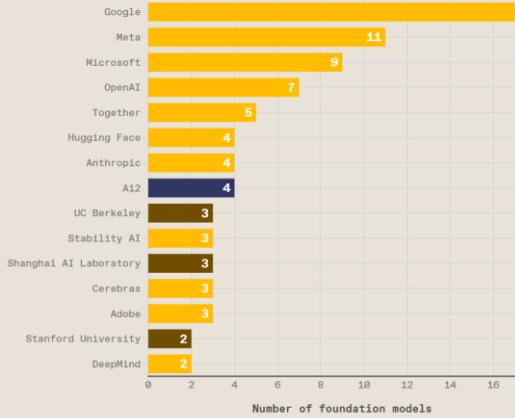
Total investment (in billions of U.S. dollars)



Generative AI investment skyrockets

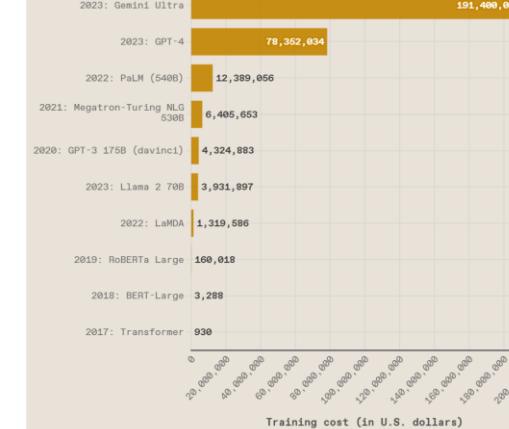
Number of Foundation Models by Organization, 2023

● Academia ● Nonprofit ○ Industry



Google is dominating the foundation model race

Estimated training cost of select AI models, 2017–23



Foundational models have gotten super expensive

Employment of new AI PhDs in the United States and Canada by sector, 2010–22

● Industry ● Government ○ Academia

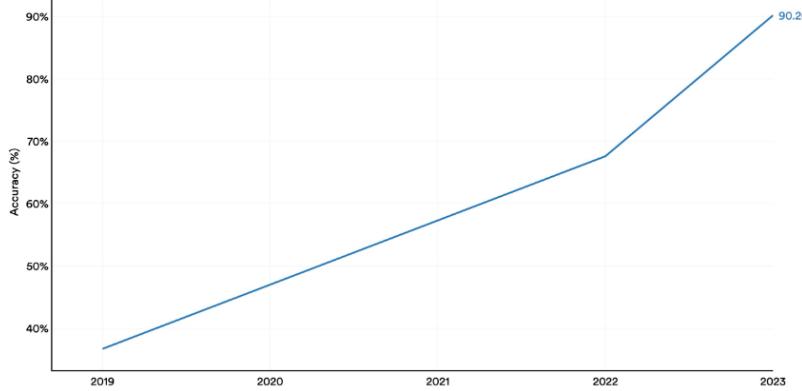
Number of new AI PhD graduates



Industry calls new PhDs

MedQA: accuracy

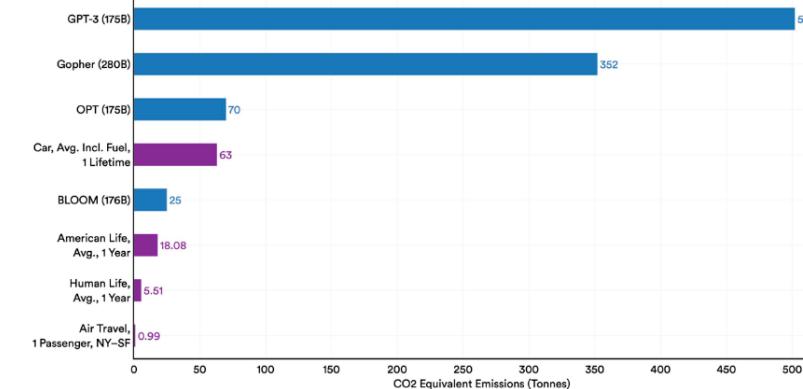
Source: Papers With Code, 2023 | Chart: 2024 AI Index report



GPT-4 Medprompt, a stand-out model of GPR-4 has reached an accuracy rate of 90.2% in clinical knowledge

CO<sub>2</sub> Equivalent Emissions (Tonnes) by Selected Machine Learning Models and Real Life Examples, 2022

Source: Lucioni et al., 2022; Strubell et al., 2019 | Chart: 2023 AI Index Report



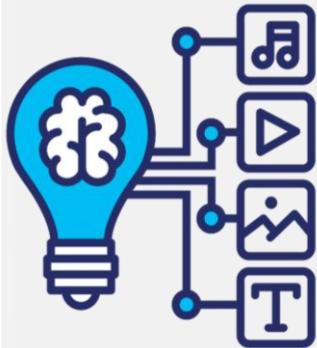
AI systems can have serious environmental impacts!

Source: Stanford's 2024 AI Index Tracks Generative AI and More - IEEE Spectrum



# State of the Art in AI

## Generative AI



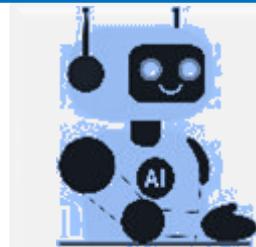
- **LLMs:** OpenAI's GPT-4, Google's PaLM, Meta's LLaMA, and Anthropic's Claude are being integrated into various applications such as chatbots, content creation, coding assistance, and more.
- **Multimodal Models:** OpenAI's GPT-4 Vision and DeepMind's Gemini combine language, vision, and other modalities can analyze images, generate descriptions, and even perform image editing.
- **Diffusion Models:** Diffusion models like DALL-E 3, Midjourney, and Stable Diffusion have set new standards for generating high-quality, realistic image, video, and audio content

## RL



- **Real-World Applications:** RL is being applied to robotics like iRobot's Roomba vacuum cleaner, autonomous driving, supply chain optimization, and personalized recommendations, where agents learn optimal strategies by interacting with their environments.
- **Advancements in Game AI:** RL models like AlphaZero and MuZero have demonstrated the ability to learn complex strategies in games such as chess, Go, and even real-time strategy games without human data.

## XAI



- IBM's AI Explainability 360, DARPA's XAI program, SHAP (SHapley Additive exPlanations) models and frameworks help understand and interpret the decision-making processes of AI systems, essential for ensuring transparency, accountability, and trust, especially in critical domains like finance and healthcare.

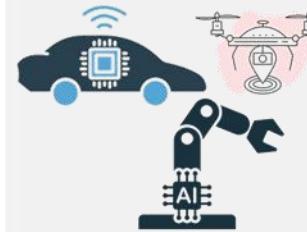
# State of the Art in AI

## AI in Healthcare



- IBM Watson for Oncology, Google's DeepMind AI for protein folding (AlphaFold), PathAI for pathology are revolutionizing healthcare by improving diagnostics, drug discovery, personalized treatment plans, and monitoring patient health.

## AI in Autonomous Systems



- Tesla's Full Self-Driving (FSD) mode, Waymo's autonomous taxis, Boston Dynamics' AI-driven robots etc. are advancing self-driving cars, drones, and robots, enabling them to perform tasks autonomously in dynamic and complex environments.

## AI for Sustainability



- Google's AI for predicting flood patterns, Climate AI for crop yield prediction, DeepMind's energy-efficient cooling for data centers address environmental challenges, including climate change mitigation, optimizing energy consumption, and predicting natural disasters

# Dangers of AI

Automation Spurred  
Job Losses/  
Displacements

Deepfakes

Privacy Violations

Social Manipulations  
Through AI Algorithms

Algorithmic Bias  
Caused By Bad Data

Market Volatility

Autonomous Weapons  
Powered by AI

Un-controllable Self-  
aware/ Sentient AI

# Challenges and Limitations

Lack of transparency

Bias and Discrimination

Privacy Concerns

Ethical Dilemma

Mis-use/ Security Risks

Concentration of Power

Potential Job Displacement

Dependence on AI

Economic Inequality

Legal and Regulatory Challenges

Loss of Human Connection

Existential Risks (AGI)

# Future Direction of AI

## Generalized and Multimodal AI Systems

- Beyond Narrow AI : Generalized Multi-modal AI models (text, image, audio, video etc.) across different domains with an underlying unified architecture
- Unified AI Models : Integration of different AI paradigms – supervised, unsupervised, reinforcement learning etc. – into a single, versatile model that can learn from diverse types of data and feedback reflecting more human-like learning capabilities

## Efficient and Scalable AI

- Model Efficiency and Optimization : Reduction in compute footprint, energy consumption etc.
- Federated and Edge AI : Models are trained on devices than on centralised servers, enhancing privacy, reducing latency, and enabling real-time decision-making

## Explainable and Interpretable AI – XAI

- Trust Worthy AI Systems - Bringing transparency and understandability of AI systems to humans through explainability and interpretability of their internal decision making process
- Causality in AI – Incorporating causal reasoning into AI model to go beyond correlation-based learning to usher in decision making based on deeper understanding of cause-and-effect relationship

## Ethics, Fairness and Responsibility

- Bias mitigation and fairness techniques to mitigate bias and enhance fairness in AI models to ensure they do not perpetuate or amplify societal inequalities



# Future Direction of AI

## Domain Specific AI Applications

- Specially fine tuned and trained models for different domains – healthcare, finance, education, manufacturing, legal, climate sciences etc.

## AI for Social Good and Human-AI Collaboration

- AI tools to address global challenges like climate change, food security, public health, disaster management etc.
- Creating AI tools that enhance the creativity, productivity, and decision-making in the fields of art, design, engineering, customer service etc.

## AI Governance and Regulation

- Establishing global standards and regulatory frameworks to govern the development and deployment of AI technologies

## Next generation AI Architectures

- Exploration of new computing paradigms - To overcome the limitations of classical computing enabling faster and more efficient processing of vast data sets - Neuromorphic Computing, Quantum AI etc.
- Self-supervised and Few-Shot Learning – Learning from unlabelled data and from few examples making AI learning process more data-efficient and adaptable to new tasks with minimal retraining





**“GAI would be a pervasive and ubiquitous API utility in each technology solution on the planet like a RESTful End-point”**

# AAM-IPL Overview

## Program benefits

- Hands-on experience in coding and implementing AI/ML algorithms using real-world industry data
- Lab will guide students through the practical application of key ML techniques
- Students will engage in project-based learning, applying AI and ML techniques they have learnt in their curriculum to address practical industry challenges
- Bridge the gap between theoretical knowledge and practical application of ML algorithms
- Invaluable insights into current industry practices and trends
- Enhancing technical proficiency for successful careers

## ML algorithms addressed

- PCA, Linear Regression, Logistic Regression
- Support Vector Machines, K-means clustering, Naive Bayes
- Multi-variate Gaussian Discriminant Analysis, Decision Tree Classification

## Pre-requisites

- Completion/studying ML course of CSE(AIML) scheme 2020
- Students are expected to be familiar with basic concepts of
  - Linear Algebra, Properties and Matrices operations
  - Calculus – differential, integration
  - Convex Optimization etc.



# AAM-IPL Overview

## Structure of the Course

- Each Monday 8:00 PM IST
- Next Project Announcement – Topic, Data Set, Shell Code
- Announcement Channels – Google Classroom, Industry Projects WhatsApp Group
- Each Saturday 10:30 AM IST – 12:00 Noon IST
- Overview of the algorithm and project, interactive Q&A on the project implementation
- Each Sunday 10:30 AM IST – 12:00 Noon IST
- Project implementation, output demonstration, interactive Q&A
- Each Sunday 11:59 PM IST
- Deadline to submit/upload completed week's project code onto Google classroom
- All online classes will be through Google Meet (provided by Brillium Technologies)
- Project submissions will be graded

## Tools and Technologies

- Computing Language – Python
- IDE – Visual Studio Code with Jupyter Notebook
- Online Data Sets for Training - Datacamp, Kaggle, Wikipedia

## Online Google Classroom

- FCH, presentation slides, projects data, problem statement will be uploaded to AAM – IPL [Google Classroom](#) ahead of the respective sessions
- Program schedule is available in [Google Classroom](#)

## Program and Lab Coordinator

- Prof. V. Suresh - [vsuresh.ecs@gprec.ac.in](mailto:vsuresh.ecs@gprec.ac.in)

## Interested in Gen AI Projects?

- Reach out to [venkat@brillium.in](mailto:venkat@brillium.in) with your idea





# Call To Action



- Asking questions – there is **No Stupid Question**
- Understanding **the algorithm** of the project **early in the week after announcement**
- Asking project implementation related **questions on each Saturday/project overview sessions**
- Implementing the project **before each Sunday session** and demonstrating the output, if asked
- Seeking help to completing the project **in Sunday sessions**, if needed
- Submitting the completed project code (runnable) **by each 11:59 PM IST on Sunday**



**Interested in building a Gen AI application?  
Reach out to [venkat@brillium.in](mailto:venkat@brillium.in)**



## Applied AI & ML Industry Projects Lab (AAM-IPL)

Transforming Learning into Industry Solutions

# THANK YOU!

AAML-IPL Brought You in Partnership with:



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