





Navigating the Artificial Intelligence Revolution Luís Paulo Reis

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Agenda

- Navigating the New World of Al
- The Evolution of Artificial Intelligence
- Artificial Intelligence (AI) and Machine Learning (ML)
- Generative AI and Large Language Models (LLM)
- Al Applications and Challenges
- Agents, Agentic Al and Multi-Agent Systems
- Robotics, Deep RL, LBMs and the Future
- Conclusions



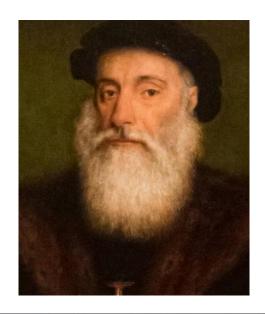
There are More Tides than Sailors







There are More Tides than Sailors

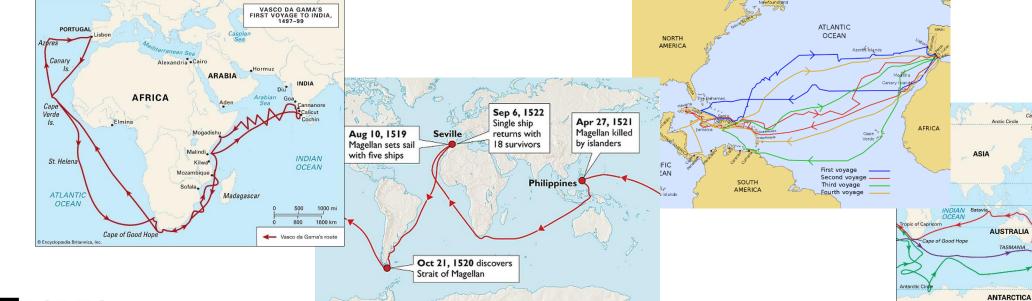








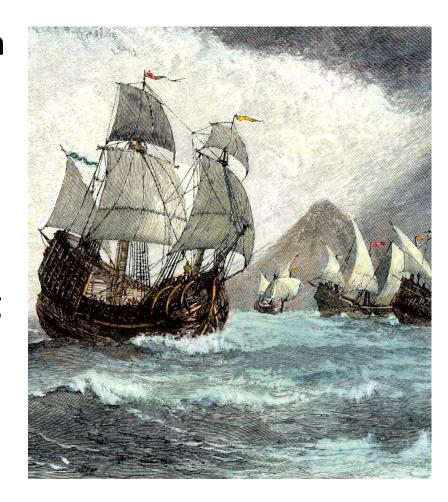
SOUTH
Archipelago AMERICA







- Ferdinand was born in Portugal in 1480
- In September 1519, Ferdinand Magellan along with 5 ships left from Spain
- He convinced the king of Spain to pay for his expedition
- Find a way to reach
 Asia by sailing west
 around the Americas





- First sailed to South America
- Tried to find a river that would take him across South America to the Pacific Ocean
- Finally in 1520 he found a kind of river:
 Magellan Strait
- For 3 months, ships sailed across the Pacific Ocean
- Began running out of food and water
- Many men died from hunger and diseases

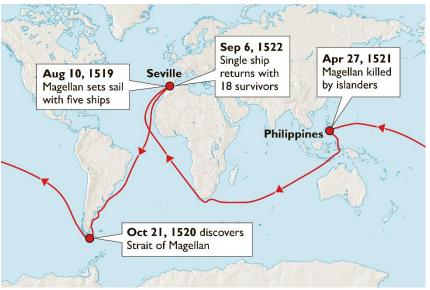






- March 1521, ships reached **Philippines**
- On April 27, Magellan was killed in battle against warriors of Lapulapu
- In 1522, only one of the five ships returned to Spain
- Of the 270 men, only 18/19 survivors returned
- First to sail completely around the world









Pacific Ocean

Stroit of Mogellon

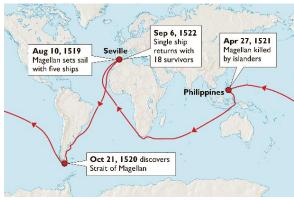
"The greatest sea voyage in the Age of Discovery!" "The most important maritime voyage ever undertaken!"

Vision,
Courage,
Leadership,
Skill and
Perseverance!

King Manuel I of Portugal did not invest in it!









Al and the Discovery of the New World





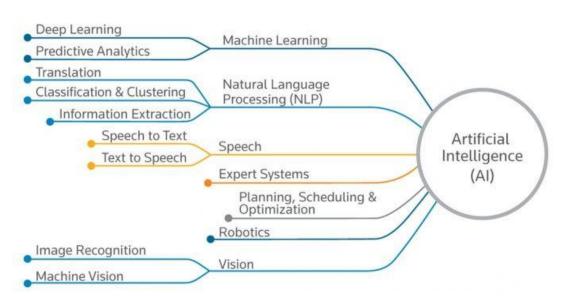
Artificial Intelligence (AI)

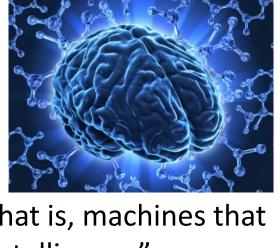
Intelligence

 "Capacity to solve new problems through the use of knowledge"

Artificial Intelligence

 "Science concerned with building intelligent machines, that is, machines that perform tasks that when performed by humans require intelligence"











Weak and Strong Al

Weak Artificial Intelligence

weak AI, also known as narrow AI is artificial intelligence that is focused on one single narrow task

Strong Artificial Intelligence

Strong AI or Artificial General Intelligence (AGI) is the intelligence of a machine that could successfully perform **any intellectual task** that a human being can! **Science fiction?** Not any more...



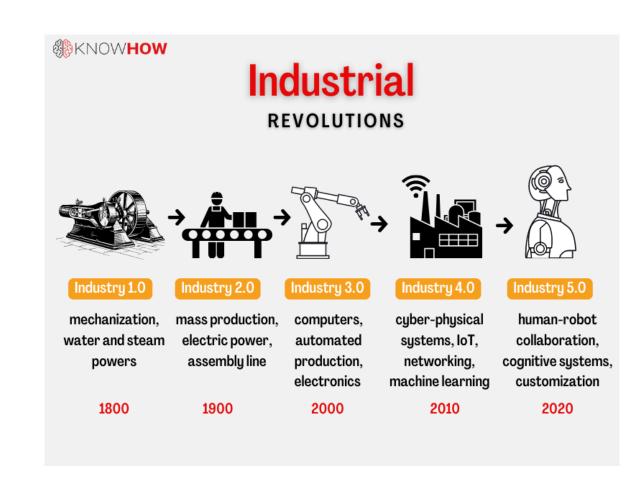
Image: https://livingsmartheart.com/what-is-strong-ai/





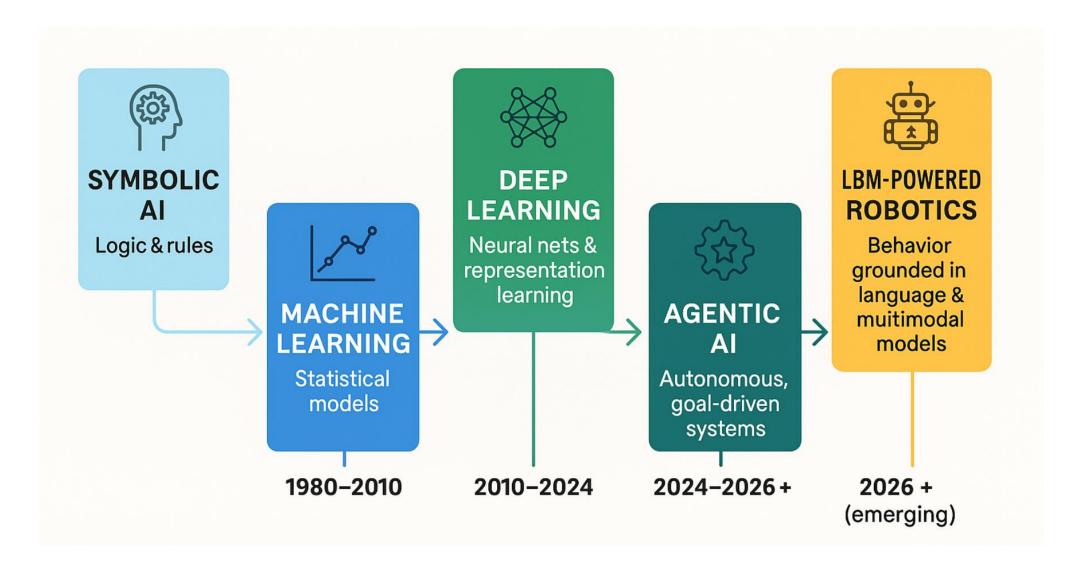
The World Is Changing Faster Than We Can Understand

- Al won't replace humans but humans using Al will replace humans not using Al
- Al is moving from tools that assist humans to agents that act autonomously
- Autonomous, Physical, Goal-Driven, Learning Systems





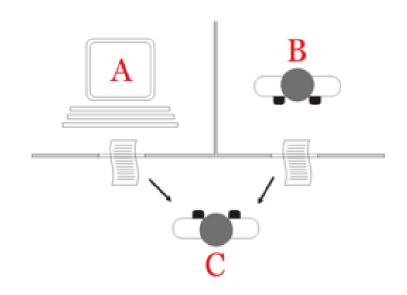
Evolution of AI Paradigms





The Birth of Al

- From Logic to Intelligence...
- The Dream Begins (1950–1980)
- "Al began as a dream of formalising reasoning!



1956 Dartmouth Conference: The Founding Fathers of AI



John MacCarthy





Marvin Minsky



Claude Shannon



Ray Solomonoff



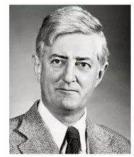
Alan Newell



Herbert Simon



Arthur Samuel



Oliver Selfridge



Nathaniel Rochester



Trenchard More



https://jlanyon.substack.com/p/a-brief-history-of-ai

AI Timeline

A.I. TIMELINE











1950

TURING TEST

Computer scientist Alan Turing proposes a test for machine intelligence. If a machine can trick humans into thinking it is human, then it has intelligence

1955

A.I. BORN

Term 'artificial intelligence' is coined by computer scientist, John McCarthy to describe "the science and engineering of making intelligent

1961

UNIMATE

First industrial robot, Unimate, goes to work at GM replacing humans on the assembly line

1964

ELIZA

Pioneering chatbot developed by Joseph Weizenbaum at MIT holds conversations with humans

1966

SHAKEY

The first electronic person' from Stanford, Shakey is a generalpurpose mobile robot that reasons about its own actions

A.I.

WINTER

Many false starts and dead-ends leave A.I. out in the cold

1997

DEEP BLUE

Deep Blue, a chessplaying computer from IBM defeats world chess emotionally intelligent champion Garry Kasparov

1998

KISMET

Cynthia Breazeal at MIT introduces KISmet, an robot insofar as it detects and responds to people's feelings



















1999

AIBO

Sony launches first consumer robot pet dog autonomous robotic AiBO (Al robot) with skills and personality that develop over time

2002

ROOMBA

First mass produced vacuum cleaner from iRobot learns to navigate interface, into the and clean homes

2011

Apple integrates Siri, an intelligent virtual assistant with a voice iPhone 4S

2011

WATSON

IBM's question answering computer Watson wins first place on popular \$1M prize television quiz show

2014

Eugene Goostman, a chatbot passes the Turing Test with a third of judges believing Eugene is human

2014

ALEXA

Amazon launches Alexa, an intelligent virtual assistant with a voice interface that completes inflammatory and shopping tasks

2016

Microsoft's chatbot Tay goes roque on social media making offensive racist

2017

ALPHAGO

Google's A.I. AlphaGo beats world champion Ke Jie in the complex board game of Go. notable for its vast

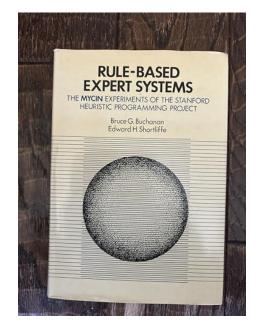
Image: [Paul Marsden, 2017] - https://digitalwellbeing.org/artificial-intelligence-timeline-infographic-from-eliza-to-tay-and-beyond/





Symbolic AI and Expert Systems

- Knowledge is Power Encoding the Mind into Rules!
- Symbolic AI represents knowledge through symbols and rules
- Examples:
 - MYCIN (medical diagnosis)
 - DENDRAL (chemical analysis)
- Architecture: inference engine + knowledge base
- Early AI was explainable and logical!
- But could not learn or adapt...







From Rules to Patterns

Al needed to evolve from being programmed to being trained

- From Rules to Patterns AI Started to Learn from Data
- Instead of telling the machine what to do, we let it learn how to do it
- Machine Learning (ML) = algorithms that learn patterns from data



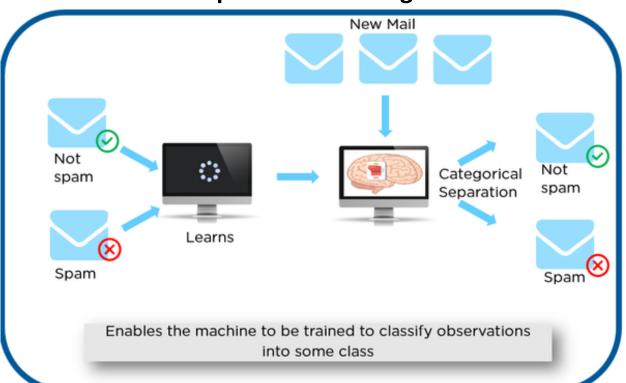
- Examples: email spam filters, handwriting recognition, early speech recognition
- ML marked the shift from explicit logic to statistical inference



Machine Learning

Field of artificial intelligence that gives **computer systems** the **ability to** "learn" (e.g., progressively **improve performance** on a specific task) from **data/results of their actions**, without being explicitly programmed

Supervised Learning



Reinforcement Learning

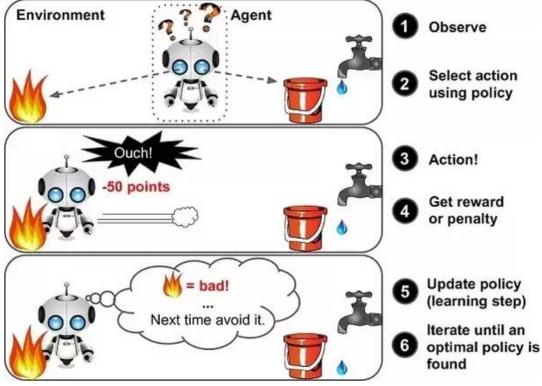


Image: https://towardsdatascience.com/what-are-the-types-of-machine-learning-e2b9e5d1756f

Image: https://becominghuman.ai/an-introduction-to-machine-learning-33a1b5d3a560

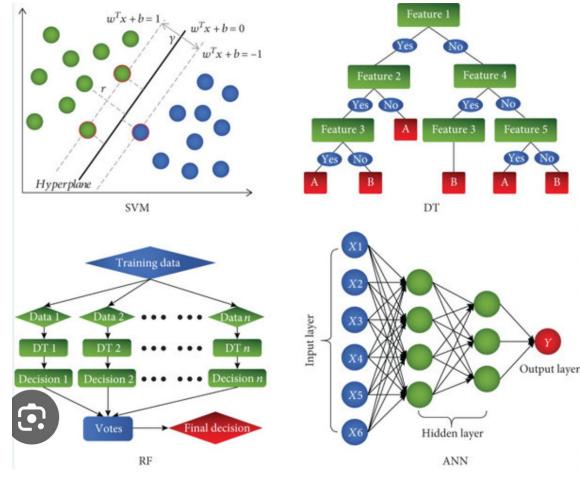




From Statistics to Intelligence

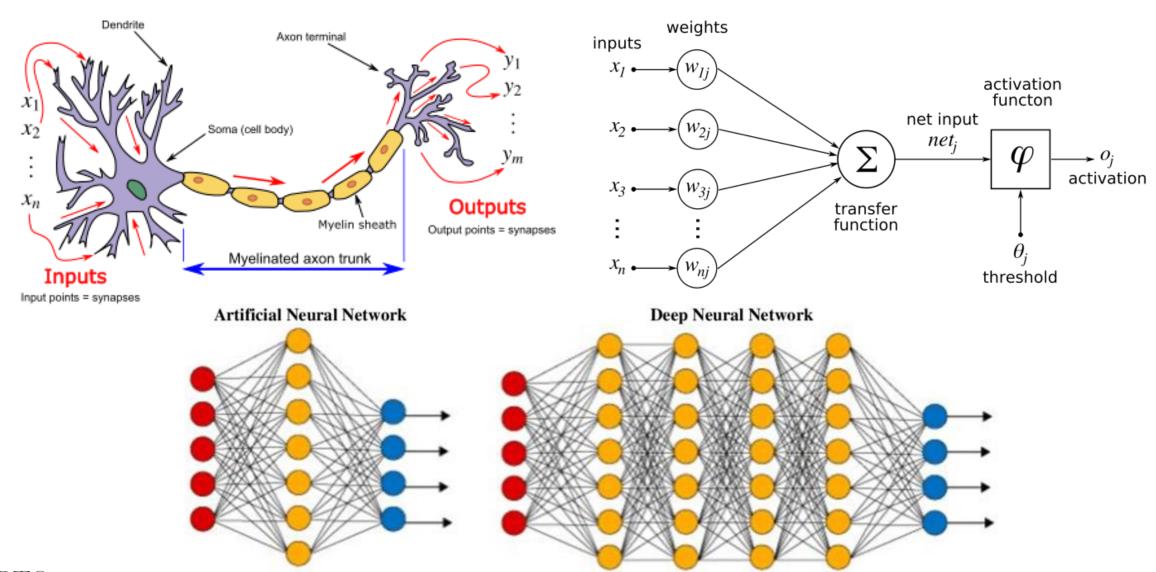
- Algorithms and Breakthroughs
- Between the 1980s and early 2000s, we saw breakthroughs:
- Decision Trees, Random Forests, SVMs, Neural Networks and the rediscovery of Backpropagation
- Paving the road to Deep Learning







Artificial Neural Networks



The Data Revolution

- Data Is the New Oil and the Fuel of AI
- With the internet, smartphones, and sensors, data grew exponentially
- Algorithms improved but what truly powered modern AI was scale!
- Algorithms got better but data made them powerful!

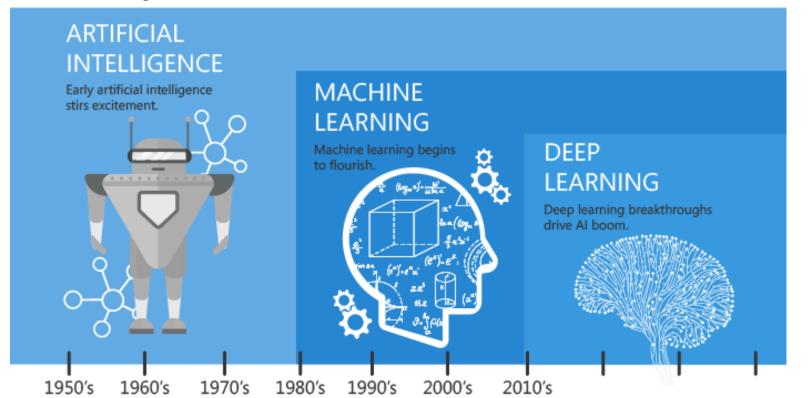


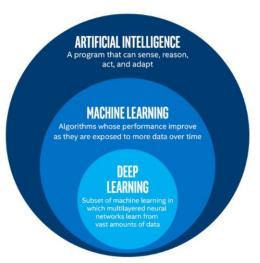


The Deep Learning Revolution

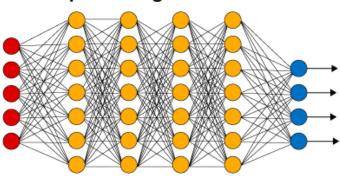
- When Neural Networks Got Deep and Everything Changed!
- AlexNet on ImageNet triggered the deep-learning boom

Multi-layer neural networks, learn features automatically





Deep Learning Neural Network



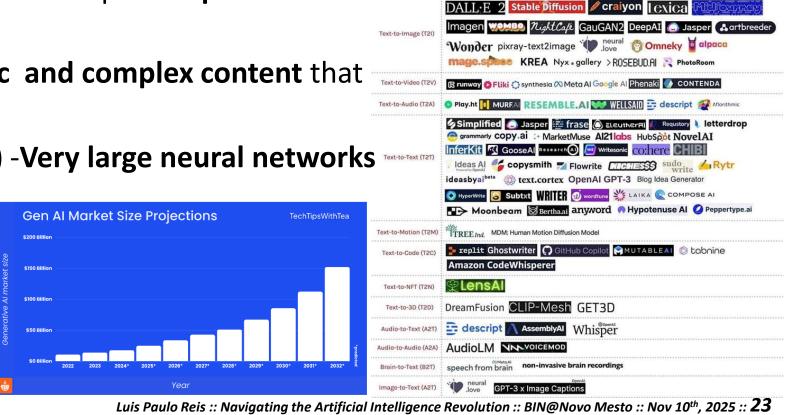


Generative AI - Machines That Create

- Generative AI (GenAI) can create a wide variety of data, such as images, videos, audio, text, and 3D models
- Learning patterns from existing data, and then using this knowledge to generate new and unique outputs in response to prompts
- GenAl produce highly realistic and complex content that mimics human creativity

Large language models (LLM) -Very large neural networks

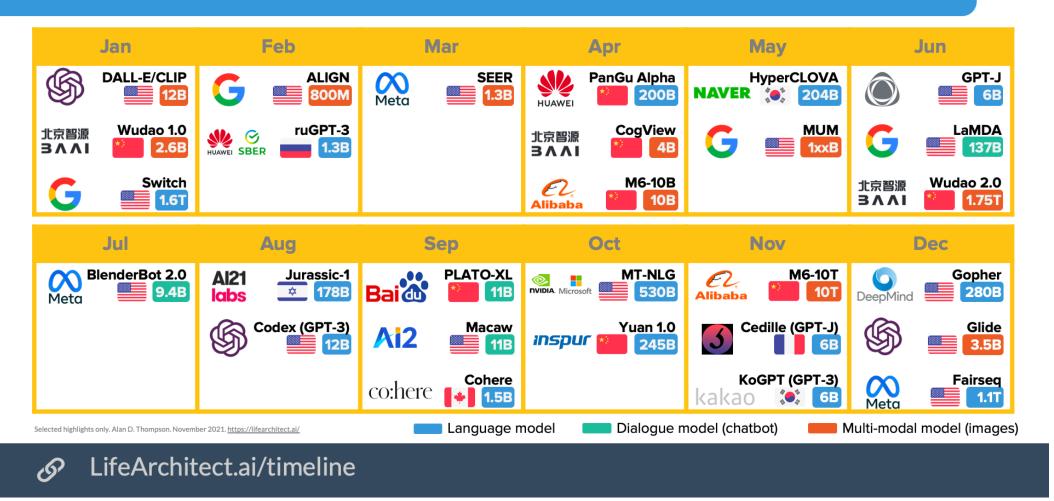
with billions of weights, trained on very large quantities of text





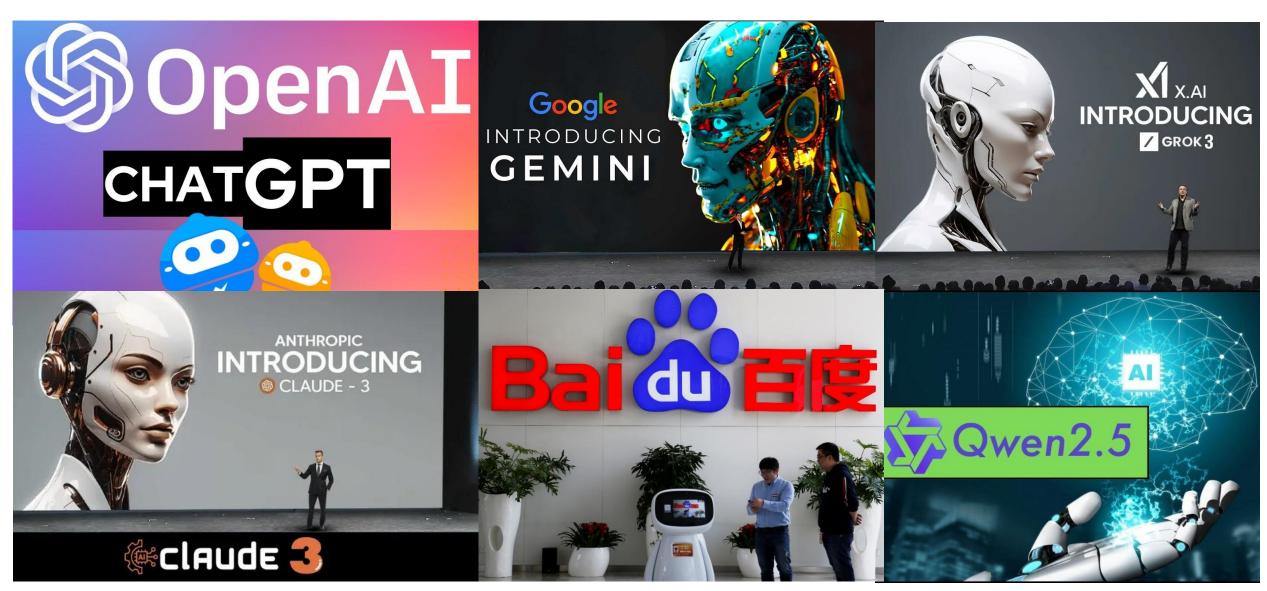
AI Timeline

AI TIMELINE: 2021 MORE THAN 24 LARGE MODELS IN LESS THAN 12 MONTHS





Generative AI Competition





LLM Evolution 2022/3->2024/5

2022-231

Anthropic Claude

- Not multimodal (text only)
- Limited contextual understanding (difficulty with complex conversations)
- No tool usage

Goodle Rard

➤ Claude 3.5

Jan 2025²

- · Multimodal (text, audio, and images)
- Enhanced contextual understanding and coherence during long interactions
- Experimental computer usage capability for some users

Comini 2 0 Flach

OpenAl

GPT-3.5

- Not multimodal (text only)
- Fair reasoning ability (eg, scored high on SAT, but bottom 10% on bar examination)
- Limited contextual understanding (difficulty with coherence in complex conversations)
- Standard API access (for text generation)

OpenAl o1

- Multimodal (text and images)
- Advanced reasoning (eg, top 10% on bar examination)
- Enhanced contextual understanding (maintains coherence in long dialogues)
- Advanced API access (supports multimodal inputs)

Google Gemini

Google Bard

- Not multimodal (text only)
- Fair reasoning
- Limited contextual understanding (difficulty with complex conversations)
- Limited real-time data integration
- Low personalization (limited adaptability)

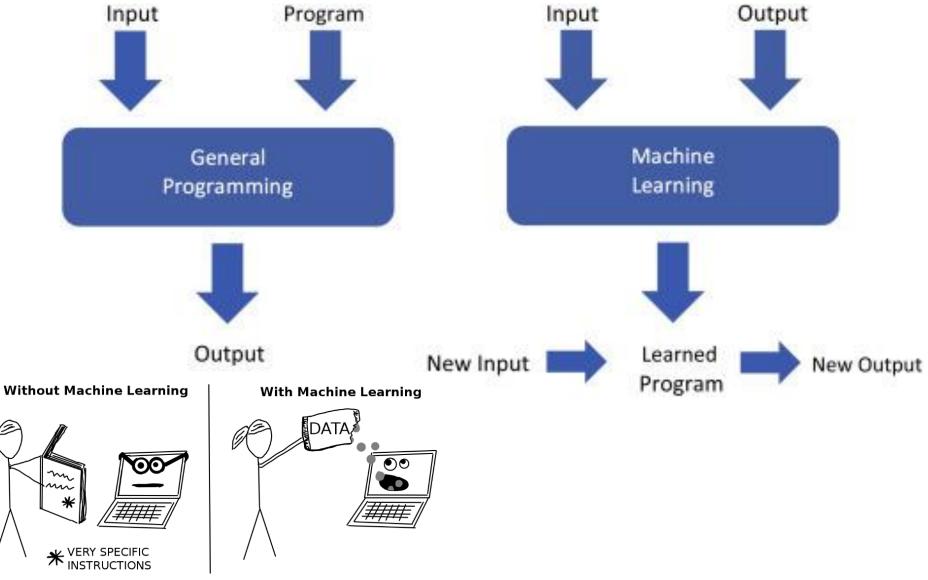
Gemini 2.0 Flash

- Multimodal (text, audio, and images)
- Advanced reasoning (capable of multistep problemsolving and nuanced analysis)
- Enhanced contextual understanding (maintains coherence in long dialogues)
- Real-time data integration (from Google Search)
- Advanced personalization (user context)
- rair reasoning ability (eg, scored nigh on SAI, but bottom 10% on bar examination)
- Limited contextual understanding (difficulty with coherence in complex conversations)
- Standard API access (for text generation)

- Advanced reasoning (eg, top 10%) on par examination)
- Enhanced contextual understanding (maintains coherence in long dialogues)
- Advanced API access (supports multimodal inputs)



Programming vs Machine Learning



The End of Programming!

Nvidia CEO predicts the death of coding — Jensen Huang says AI will do the work, so kids don't need to learn



By Benedict Collins published February 26, 2024

Jensen Huang believes coding languages are a thing of the past











(Image credit: Nvidia)

Nvidia CEO Jensen Huang has once again announced the death of coding, but this time in front of a potentially far more influential audience.

StarCoder 2 is a code-generating AI that runs on most GPUs

Kyle Wiggers @kyle_I_wiggers / 2:00 PM UTC • February 28, 2024





in Image Credits: Tippapatt / Getty Images

Developers are adopting Al-powered code generators — services like GitHub Copilot and Amazon CodeWhisperer, along with open access models such as Meta's Code Llama — at an astonishing rate. But the tools are far from ideal. Many aren't free. Others are, but only under licenses that preclude them from being used in common commercial contexts.

Perceiving the demand for alternatives, Al startup Hugging Face several years ago teamed up with ServiceNow, the workflow automation platform, to create StarCoder, an open source code generator with a less restrictive license than some of the others out there. The original came online early last year, and work has been underway on a follow-up, StarCoder 2, ever since.

StarCoder 2 isn't a single code-generating model, but rather a family. Released today, it comes in three variants, the first two of which can run on most modern consumer GPUs:

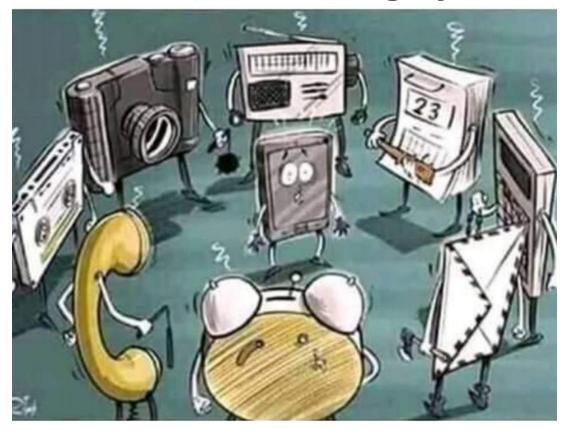
- A 3-billion-parameter (3B) model trained by ServiceNow
- A 7-billion-parameter (7B) model trained by Hugging Face
- A 15-billion-parameter (15B) model trained by Nvidia, the newest supporter of the StarCoder project



Technology: Adapt or Resist?!

- COVID-19 Use of technology
- Educate New Generations
- Ethical Use
- Technical Use
- Al Policy
- Prompt Engineering

So! You are the guy...



That took all of our Jobs!



Concerns of AI and ChatGPT in Education

- More in person exams and oral assignments
- Continuous evaluation of projects
- Help students develop their critical thinking skills by comparing Alproduced content with reliable, valid sources of information
- Be careful with over-reliance on Al
- Shift the object of evaluation from the "product" to the "process" of student learning
- Emphasize **ethical use of AI** and the importance of proper research and citation practices and avoiding plagiarism



Al and ChatGPT in Education

"Al and Can Complement Learning, Not Replace It!"
From Knowledge Transmitter → Learning Architect







Concerns of AI and ChatGPT in Research

- Risk of fabricated or inaccurate references and data ("hallucinations")
- Possible erosion of methodological rigor due to over-reliance on automated reasoning
- Challenges in ensuring reproducibility and transparency when AI is involved
- Need for explicit disclosure of Al use in research design, writing, and analysis
- Potential bias amplification from training data and opaque model behavior
- Ethical responsibility to validate, interpret, and contextualize Al outputs
 using human expertise
- Importance of maintaining authorship integrity and avoiding ghost-writing by AI



Research with Al



Scopus Al

Publish with us



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Scopus AI: Trusted content. Powered by responsible AI.

Empower your research journey with Scopus AI – your dynamic GenAI-powered research companion.

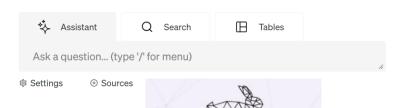
Navigate through the vast expanse of human knowledge faster with a trusted guide designed to enhance your understanding, enrich your insights, and transform your overall research experience.





Your Al Research Assistant

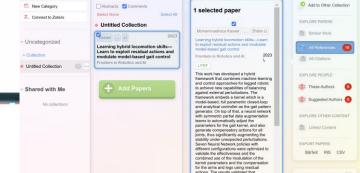
Ask questions and receive evidence-based answers from real research



Elsevier's Five Responsible Al Principles

- We consider the real-world impact of our solutions on people
- We take action to prevent the creation or reinforcement of unfair bias
- We can explain how our solutions work
- · We create accountability through human oversight
- We respect privacy and champion robust data governance









Google Search Engine

- Hummingbird Algorithm
 - PageRank algorithm that covers a specific way of giving pages credit based on the links from other pages pointing at them
 - Panda, Penguin and Payday to fight spam
 - Pigeon designed to improve local results
 - Top Heavy designed to demote ad-heavy pages,
 - Mobile Friendly designed to reward mobile-friendly pages
 - Pirate designed to fight copyright infringement
- 200 major ranking signals (up to 10,000 variations or sub-signals)
- Google RankBrain Algorithm
 - Machine-Learning System
 - 3rd most important signal
 - Interpret searches people submit to find pages that might not have the exact words that were searched for



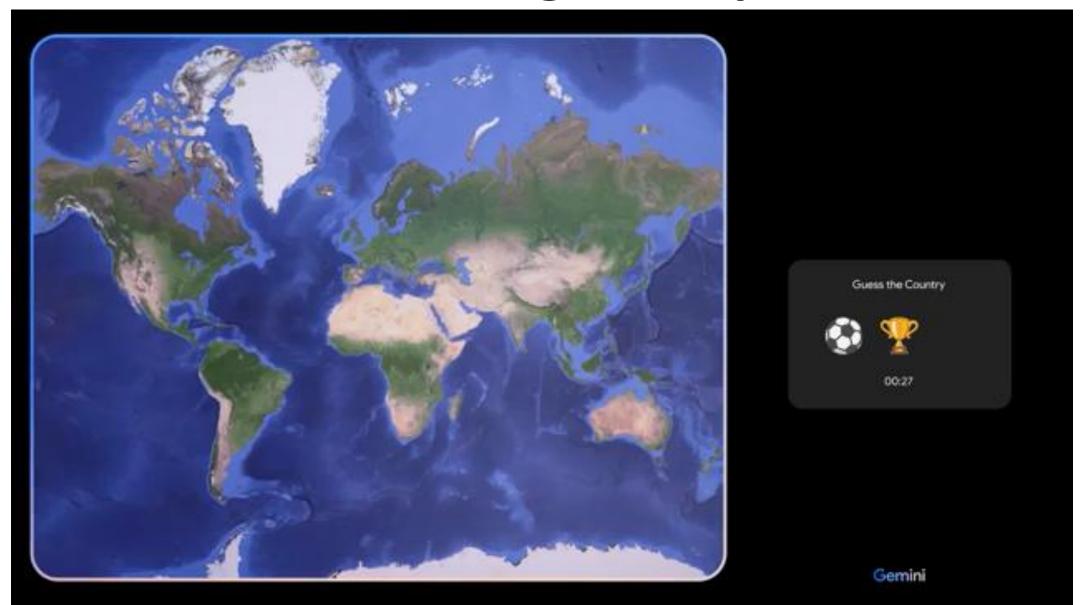






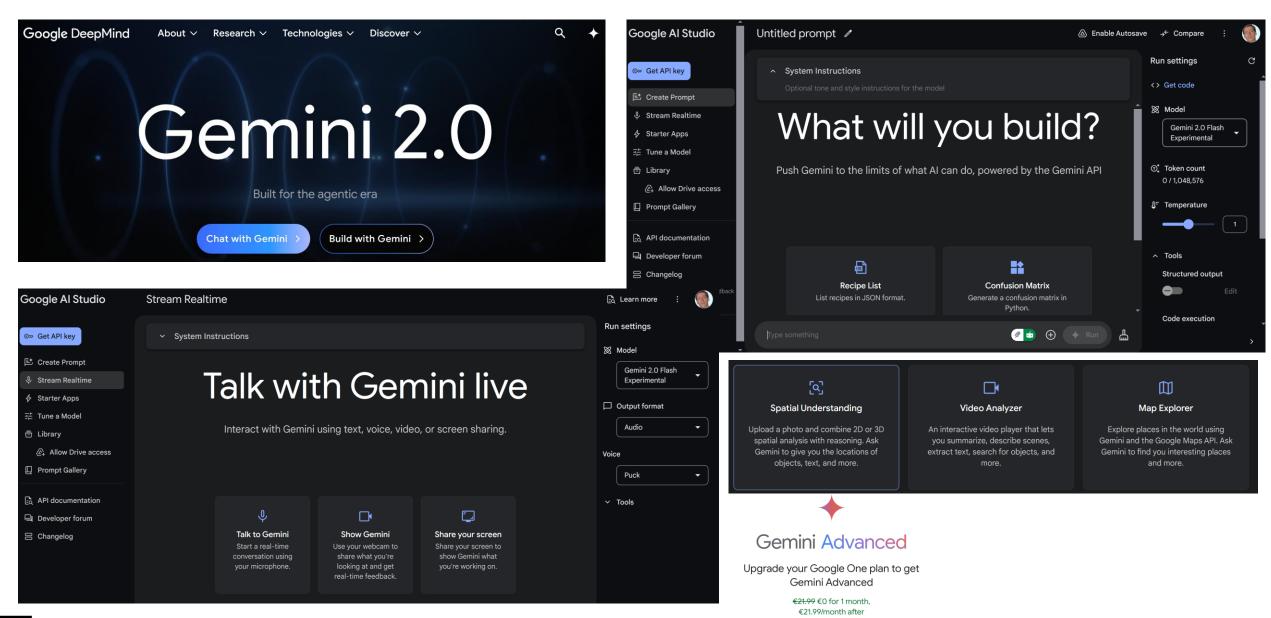


Gemini – Google Deep Mind





Gemini 2.5



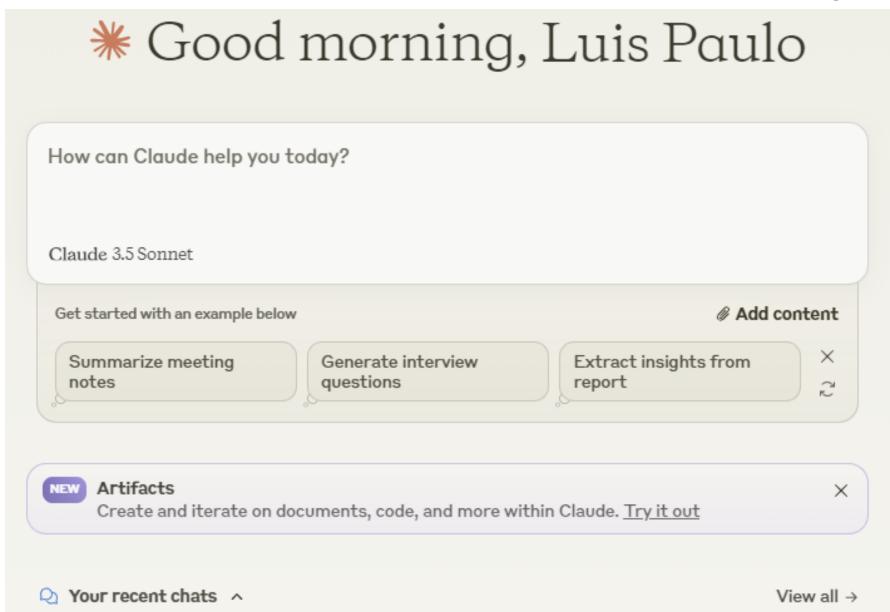


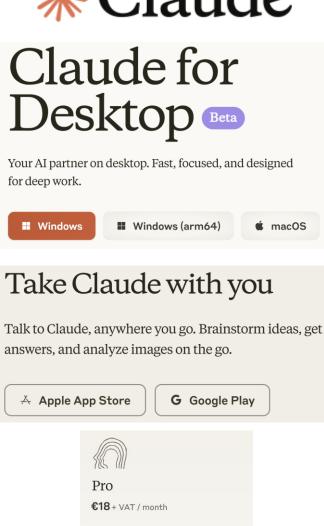
Microsoft Copilot

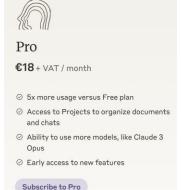


Claude 4.5 - Anthropic





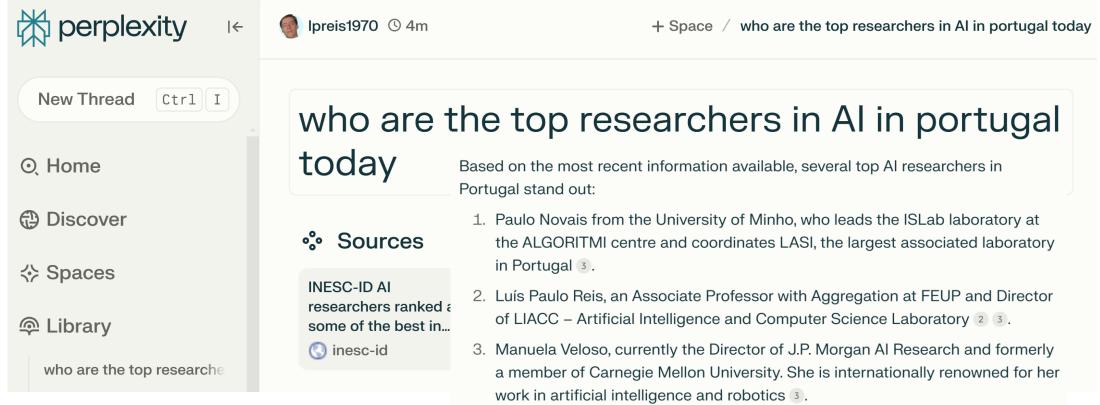




Perplexity AI

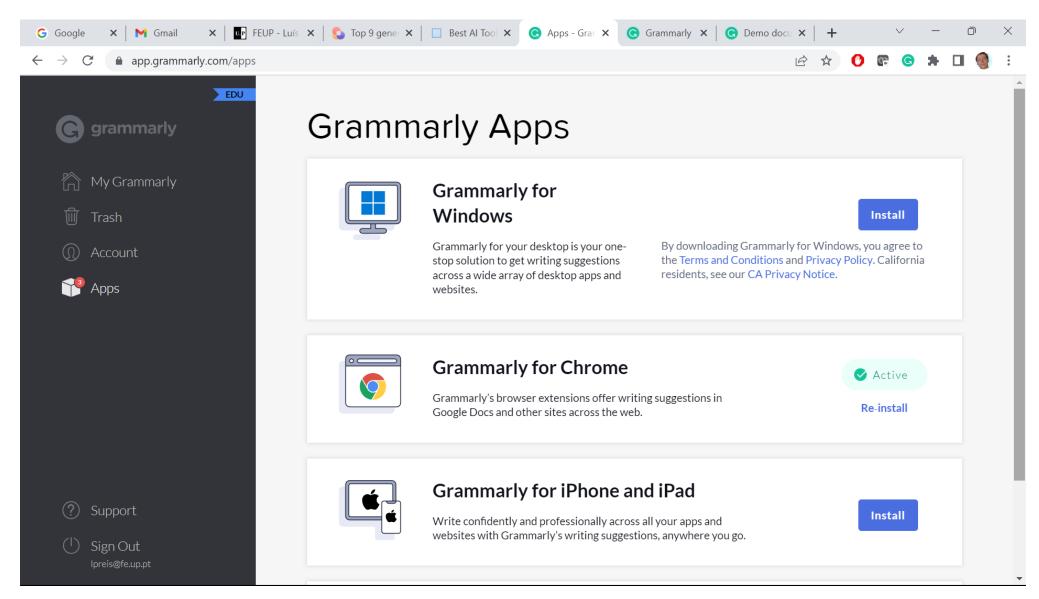


 Perplexity AI is a conversational search engine that uses large language models (LLMs) to answer queries using sources from the web and cites links within the text response.



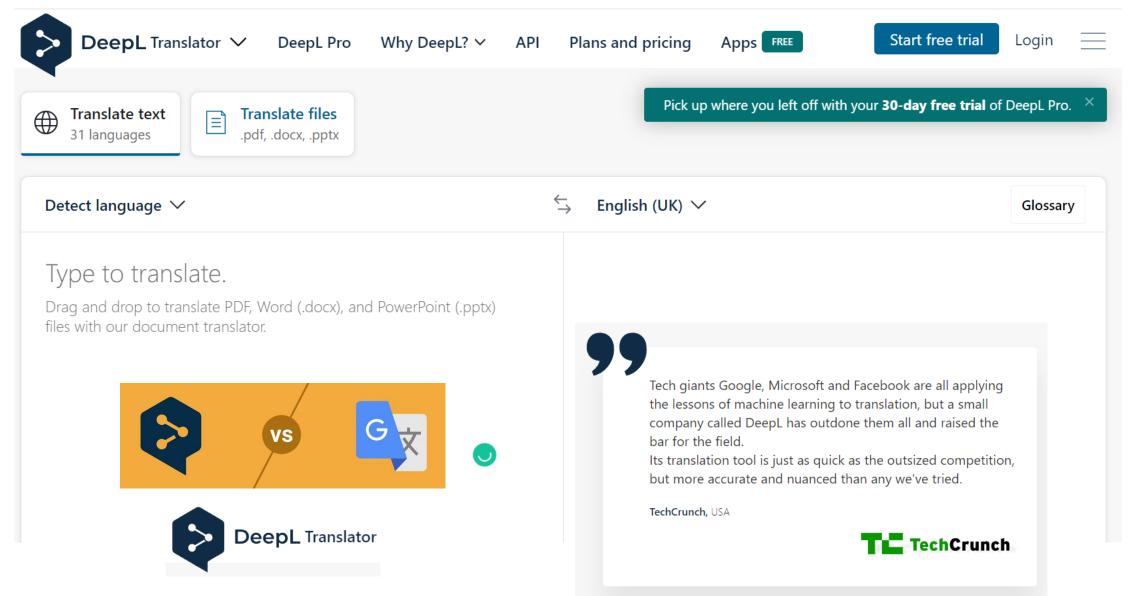


Grammarly





DeepL







QuillBot

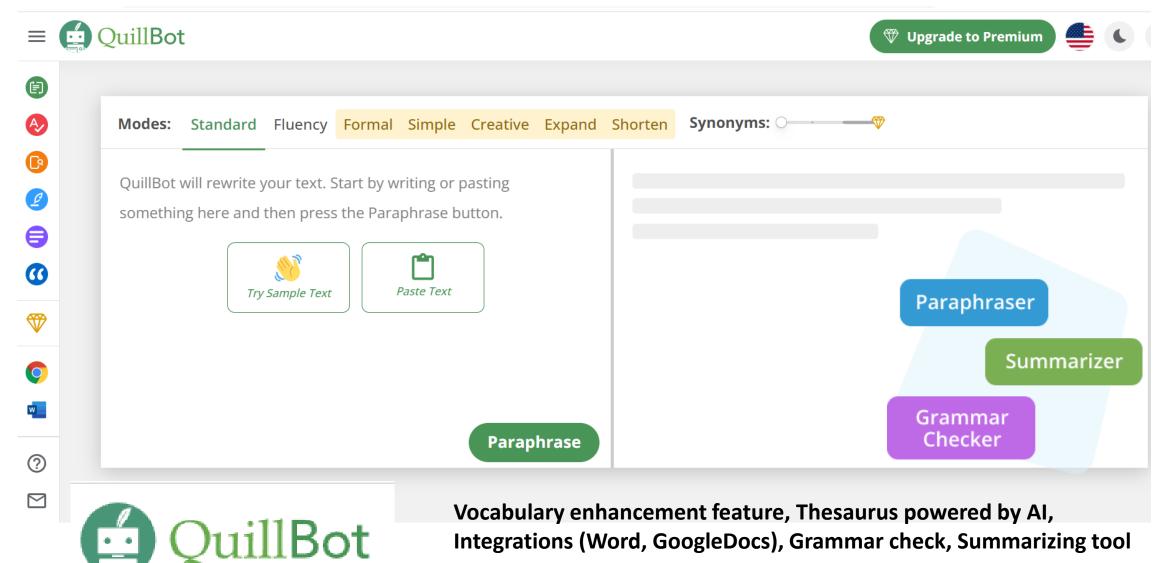




Image Generation - Dall-E 3

DALL·E 3 is an AI system that can create realistic images and art from a description in natural language

An astronaut riding a horse in photorealistic style



A beach similar to "praia da baia" in Espinho with surfers, surfing in the sea and big waves













Explore GPTs

GPTs

Discover and create custom versions of ChatGPT that combine instructions, extra knowledge, and any combination of skills.

Q Search public GPTs



Canva

Effortlessly design anything: presentations, logos, social media posts and more.

By canva.com



Consensus

Your AI Research Assistant. Search 200M academic papers from Consensus, get science-based answers, and draft content...

By Consensus



Video GPT by VEED

Al Video Maker. Generate videos for social media - YouTube, Instagram, TikTok and more! Free text to video & speech tool wit...

By veed.io



Grimoire

Logo Creator

By Chase Lean

image generator

By NAIF J ALOTAIBI

designs and app icons!

Code Wizard M Learn to Prompt-gral
Create a website (or anything) with a
sentence. 20+ Hotkeys for coding flows....

By gptavern.mindgoblinstudios.com

Use me to generate professional logo

A GPT specialized in generating and

and friendly tone.image generator

refining images with a mix of profession

Custom GPTs — special-purpose versions of the model created by others

uma receita de bacalhau com espinafres e batatascom um emprtamento bonito e foto

Talked to api.adzedek.com ∨

Receita de Bacalhau com Espinafres e Batatas

Esta é uma receita clássica e nutritiva que combina o sabor rico do bacalhau com a suavidade das batatas e a frescura dos espinafres. O prato é montado de forma elegante, com camadas visuais para um empratamento atraente.

Ingra 🤧 es:

- 400g de bacalhau dessalgado
- 300g de batatas
- 200g de espinafres frescos
- 2 dentes de alho picados
- 1 cebola em rodelas finas
- 100ml de azeite de oliva
- Sal e pimenta a gosto





Llama 4 – Meta Open Source Models

Meta

Technology ∨

Get started v

Trust & Safety >

Community >

Try Meta Al 7

Download models

Q

Build the future of Al with Meta (XX) LLaMA4

Now available with both 8B and 70B pretrained and instructiontuned versions to support a wide range of applications

Get Started

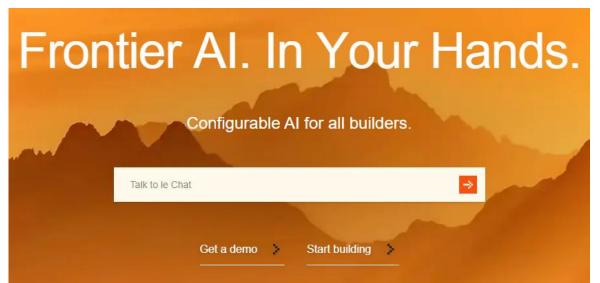
Experience Llama 3 on Meta Al

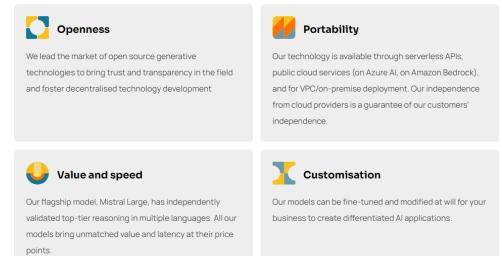






Mistral Al – Open Source Models







Build on Open Source

Under the Apache 2.0 license, our 3 open source models Mistral 7B, Mixtral 8x7B, Mixtral 8x22B are usable and customisable for a variety of use cases. They can be downloaded or used on demand via our platform.



- Download them for deployment in your own environment
- Use them on <u>La Plateforme</u> at market-leading availability, speed, and quality control



Gemma 3n – Gemini Google Open Source

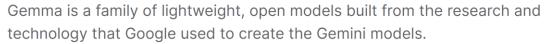
Gemma

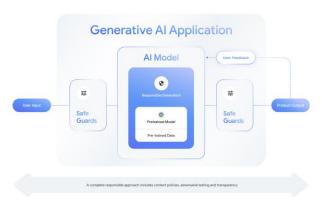
Gemma 2 is now available to researchers and developers

google/gemma

Gemma 2 offers best-in-class performance, runs at incredible speed across different hardware and easily integrates with other Al tools.

Jun 27, 2024 · 4 min read





Responsible Generative Al Toolkit

This toolkit provides resources to apply best practices for responsible use of open models such as the Gemma models, including:

- · Guidance on setting safety policies, safety tuning, safety classifiers and model evaluation.
- The Learning Interpretability Tool (LIT) for investigating Gemma's behavior and addressing potential issues.
- · A methodology for building robust safety classifiers with minimal examples.

This version of the toolkit focuses on English text-to-text models only. You can provide feedback to make this toolkit more helpful through the feedback mechanism link at the bottom of the page.

When building with Gemma, you should take a holistic approach to responsibility and consider all the possible challenges at the application and model levels. This toolkit covers risk and mitigation

Gemma 3n

Our powerful and efficient open model designed to run locally on phones, tablets, and laptops.



Send feedback



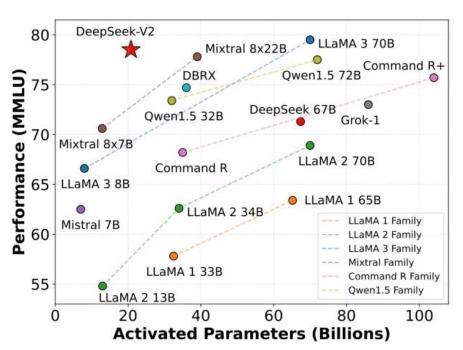




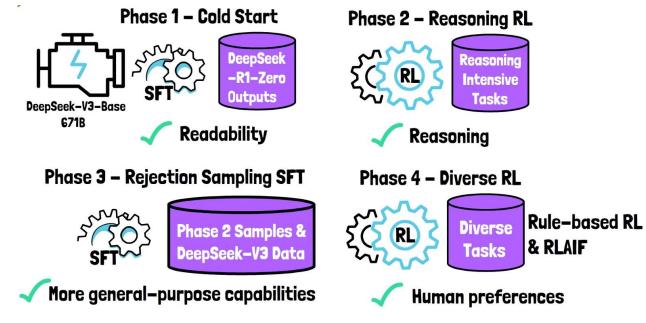


DeepSeek





DeepSeek is a Chinese AI research lab focused on developing LLMs and AI systems. DeepSeek series includes DeepSeek-V2, a high-performance open-weight model trained on diverse multilingual datasets, aiming to compete with top-tier models like GPT and LLaMA. DeepSeek emphasizes efficiency, scalability, and domain-specific capabilities, particularly in Chinese language understanding and applications. The lab has also developed DeepSeekCoder, an advanced AI model optimized for code generation and programming assistance. With strong computational resources and a growing presence in the AI research landscape, DeepSeek is positioning itself as a significant player in the global LLM ecosystem.

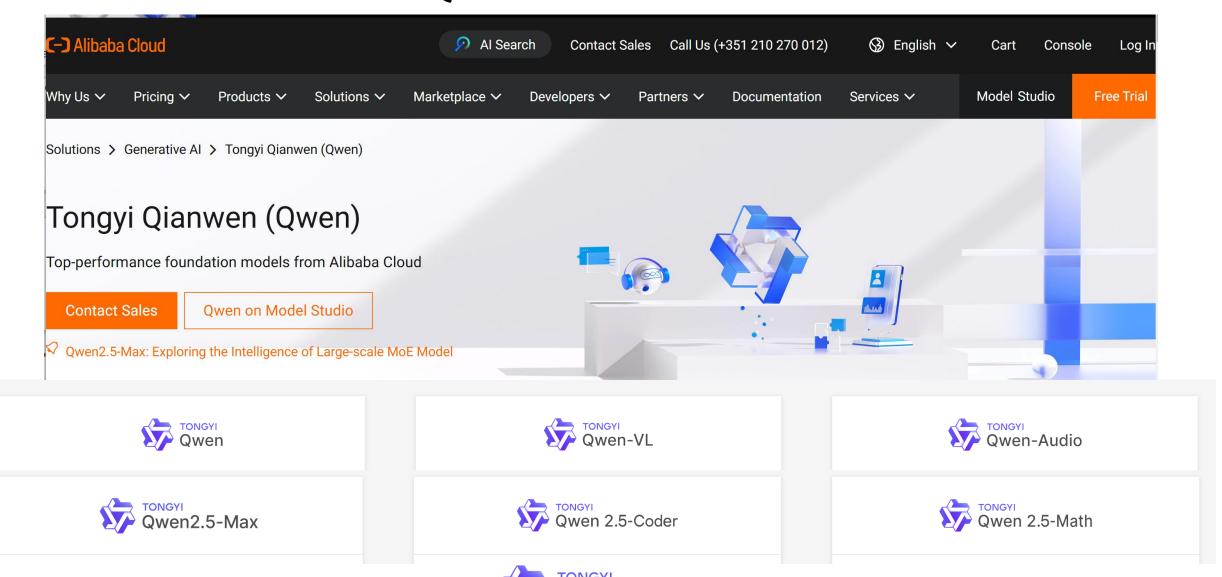






Qwen - Alibaba



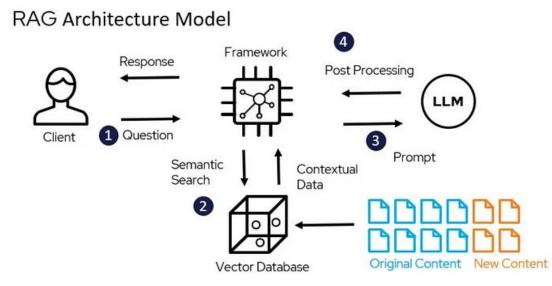


Qwen-Agent



RAG Architecture Model

- Retrieval Augmented Generation (RAG) has emerged as a powerful technique for improving LLMs
- Retrieving and conditioning external knowledge, RAG allows models to generate more accurate, relevant, and comprehensive text



https://medium.com/@gulernilay088/

 Advanced RAG enhances each module further with innovations like higher-order retrievers, cross-encoder rerankers, and evidence manipulation architectures



Al Applications in Health

- Radiology (X-ray, CT, MRI)
- Dermatology (Image)
- Drug/Treatment Discovery
- Risk Identification in Patients
- Primary Care and Screening
- Health Monitoring/Wearables
- Cognitive and Social Rehabilitation
- Physical Rehabilitation
- Patient Interaction with the Health System
- Health Systems Exchange of Information
- Surgical/Medical Robots
- Efficient Resource Allocation in Health













AI&Robotics Applications in Surgery

da Vinci Surgical System



Mako SmartRobotics



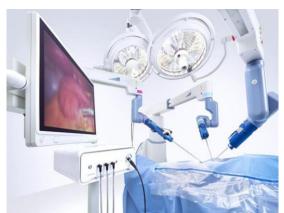
Rosa Brain Robot



Hugo Surgical Robot



Senhance Surgical System



Versisus Surgical System







Al&Robotics in Nursing

Al in Nursing:

- Personalized Treatment Plans based on individual patient data, optimizing care
- Workflow Optimization with automatization of administrative tasks, efficient scheduling, allowing nurse staff to focus more on patient care
- Patient Monitoring with Al-powered devices continuously monitor vital signs and alert nurses to any deviations
- Robotics and Automated assistants helping with physical tasks, reducing strain and increasing efficiency in patient handling and monitoring







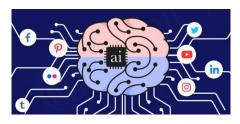


Artificial Intelligence in Digital Media

- Digital Images
- Digital Video
- Digital Audio
- Movie Industry
- Video Games
- Web Pages and Websites
- Social Media
- Digital Data and Databases
- Electronic Documents/Books







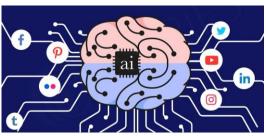






Artificial Intelligence in Social Media

- Social Creation and Management
- Social Insights
- Social Media Advertising
- Image/Face Recognition
- Fake News/Improper Content Recognition
- Personalised User Experience
- Advise Contacts/ Job candidates / Matches
- Al Powered Chatbots
- Social Listening
- Sentiment Analysis
- Improved Influencer strategies
- Increased Security











Al for the Environment

Al for Land/Agriculture

- Soil pollution, population growth, intensive agriculture
- Land cover mapping, automating and optimizing farming processes

Al for Air/Pollution

- Pollution/air quality sources identification, reduce emissions
- Intelligent and green transportation, autonomous cars, ride shares, dynamic bus routing, intelligent traffic lights

Al for Water

- Garbage collection, Intelligent ocean farming (fish and shellfish)
- Predict storms, tsunamis, hurricanes, cost-effective water management

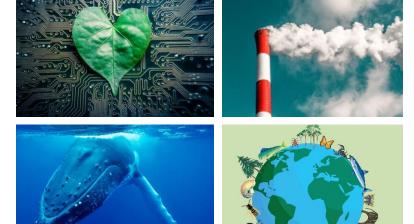
Al for Biodiversity

Surveillance, tracking, analysis, prediction

Al for Climate Change

 Better climate predictions, effects of extreme weather, predict where carbon is coming from









AI & LLM Key Trends in 2025

- Small Language Models (SLM) drive enterprise Al adoption
- Mixture of Experts (MoE) + LoRA enable SLMs to outperform 10x larger LLMs
- Open-source are becoming the de facto way for LLMs use
- Large language models start adopting modular architectures
- Synthetic data revolution arrives to LLMs
- LLM hallucinations disappear as training techniques evolve
- Data (not LLMs) becomes the true competitive advantage
- RAGs (Retrieval-Augmented Generation) used in most LLMs tasks
- Shift from closed to open-source democratizes AI (transparency)
- Agentic AI AI that acts!
- LLMs for Machines and Robots, not just people LBMs

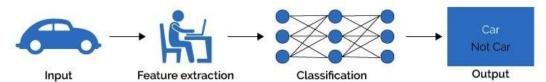


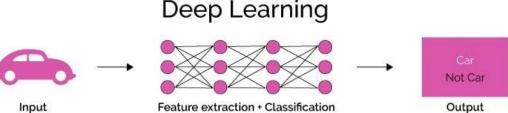
Limits of Deep Learning

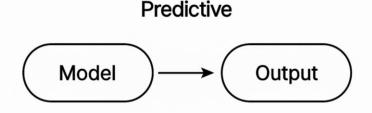
Machine Learning

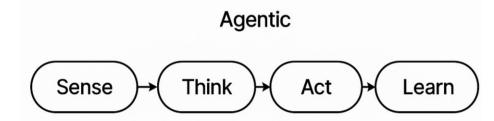
- Brilliant but Passive!
- Deep nets can predict, but they can not plan. They lack goals, intents, and common sense
- This sets the stage for Agentic AI:

 From Prediction to Decision and Action...
- The next leap is agency AI that senses, reasons, and acts toward goals
- Predictive (Model→Output) vs
 Agentic (Sense→Think→Act→Learn)











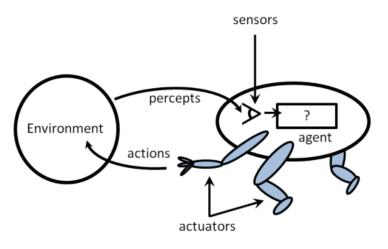
Autonomous Agents and Multi-Agent Systems

Agent (1995):

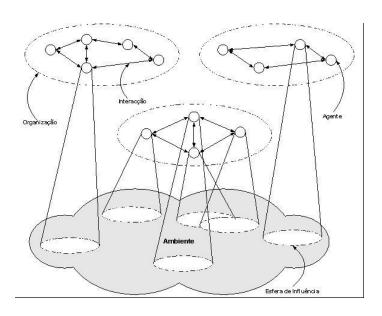
"Computational System, situated in a given environment, that has the ability to perceive that environment using sensors and act, in an autonomous way, in that environment using its actuators to fulfill a given function."

Multi-Agent System (1995):

- Agents exhibit autonomous behavior
- Interact with other agents in the system



From Russel and Norvig, "AI: A Modern Approach", 1995





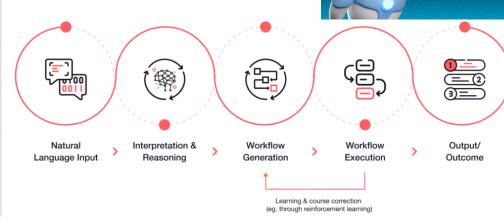
Agentic Al

- Autonomy: Ability to initiate and complete tasks with limited direct human supervision, with great flexibility and efficiency in executing tasks
- Reasoning: Sophisticated decision-making based on context and trade-offs
- Reinforcement Learning: Dynamically evolve by interacting with their environment and receiving feedback from these interactions.

Language Understanding: Capacity to understand and follow complex instructions.

• Workflow Optimization: Efficient execution of multi-step complex processes

Feature	Agentic Al	Generative AI	Traditional AI
Primary Function	Goal-oriented action & decision-making	Content generation (text, code, images, etc.)	Focused on automating repetitive tasks
Autonomy	High – Operates with minimal human oversight	Variable – May require user prompts or guidance	Low – Relies on specific algorithms and set rules
Learning	Reinforced Learning – Improves through experience	Data-driven learning – Learns from existing data	Relies on predefined rules and human intervention



AGENTIC AI



https://aisera.com/blog/agentic-ai/

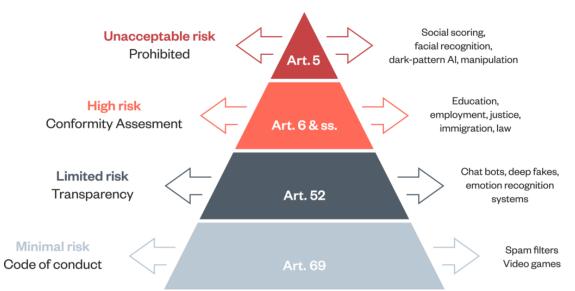
New & Emerging Trends (2025+)

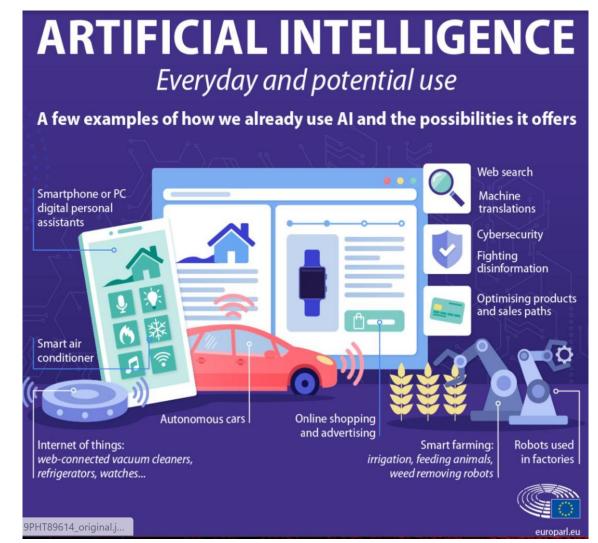
- Agentic Al Rises Autonomous agents that reason, plan, and act toward goals
- Large Behavior Models (LBMs) Models learning from actions and interactions, not just text
- Multi-Agent Collaboration Intelligent agents cooperate and negotiate to solve complex tasks
- Responsible & Regulated AI Transparency, fairness, and compliance standards
- Edge & On-Device AI Efficient, private models running locally on devices /robots
- Hybrid Human–Al Teams Shared intelligence combining human oversight and machine efficiency



EU Al Act – Regulating the Past...

- First regulation on artificial intelligence
- Al can create many benefits:
 - better healthcare
 - safer and cleaner transport
 - more efficient manufacturing
 - cheaper and more sustainable energy.







The Rise of Deep Reinforcement Learning (DRL)

Beyond Short-Term Thinking

- LLMs have limited context windows they forget.
- Memory and planning enable continuity and strategic reasoning

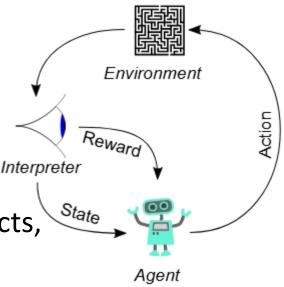
Reflection and Self-Improvement

- Evaluate their own reasoning, detect failures, and adapt
- Self-reflection transforms agents from reactive to self-correcting
- Tools/Actuators Use and World Interaction
 - Intelligence changes into capability

Learning by Doing

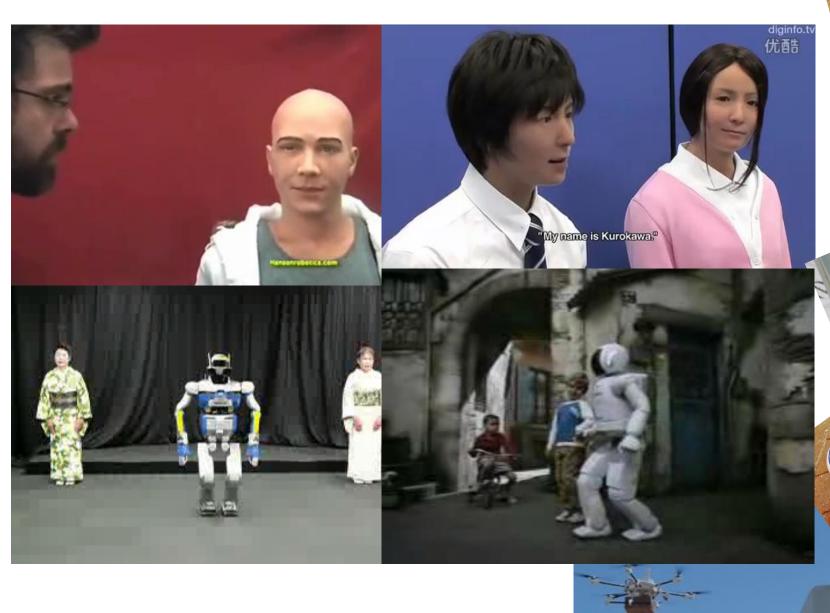
Reinforcement Learning teaches through feedback — an agent acts,
 receives a reward, and updates its policy to improve next time.







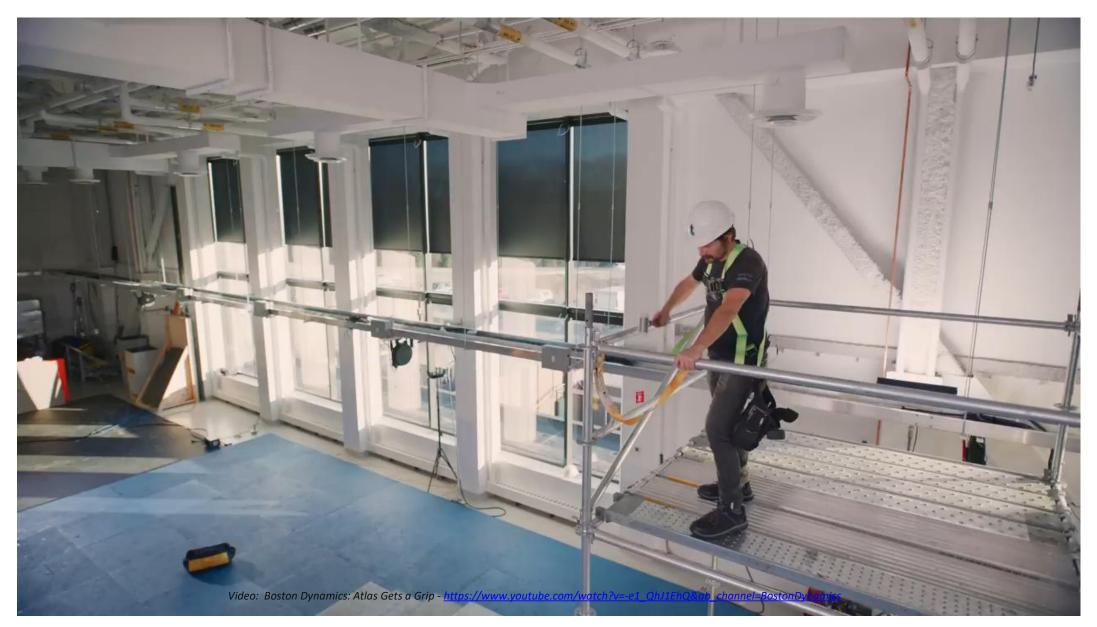
AI in Robotics





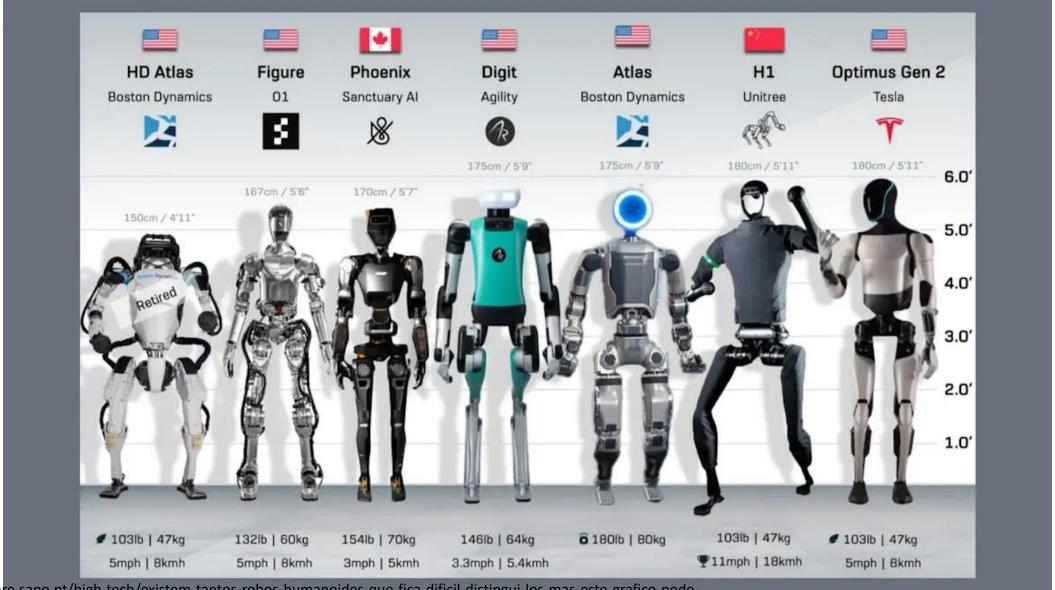


Robotics – Atlas Robot





Robotic Humanoids



https://pplware.sapo.pt/high-tech/existem-tantos-robos-humanoides-que-fica-dificil-distingui-los-mas-este-grafico-pode-ajudar/?fbclid=lwY2xjawI1Zg9leHRuA2FlbQIxMQABHS_EdekrvewLEgIyq3RZRfPy-HaU4MdNtaFgqa6bT2lkcoYfBoF_Jkqd9w_aem_RO6nDQ16bQJ51jD6bM6-TQ

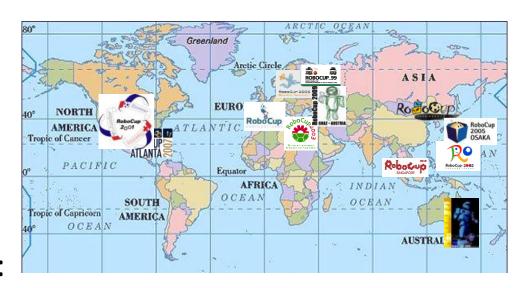


Unitree and Booster Robots



RoboCup: Objectives

- Joint International Project:
 - (Distributed) Artificial Intelligence
 - Intelligent Robotics
- Soccer Central Research Topic:
 - Very complex collective game
 - Huge number of technologies involved:
 - Autonomous Agents, Multi-Robot Systems, Cooperation, Communication, Strategic Reasoning, Robotics, Sensor Fusion, Real-Time Reasoning, Machine Learning, etc
- 5 Challenges: Soccer, Rescue, @Home, Industrial, Junior
- 17 Leagues, 2500 Humans and 2000 Robots!
- Main Goal: "By 2050, develop a team of fully autonomous humanoid robots that may win against the human world champion team in soccer!"

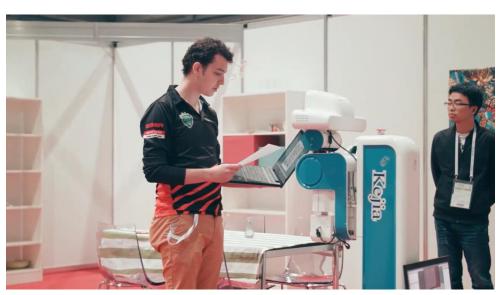


RoboCup Leagues and Challenges





Rescue





Soccer

@Home

Industrial



RoboCup - Robotic Soccer Competitions



Standard Platform League

Humanoid League

Middle

League

Small

Size

Size

RoboCup Leagues and Participations FC Portugal Team (FEUP/UAveiro)

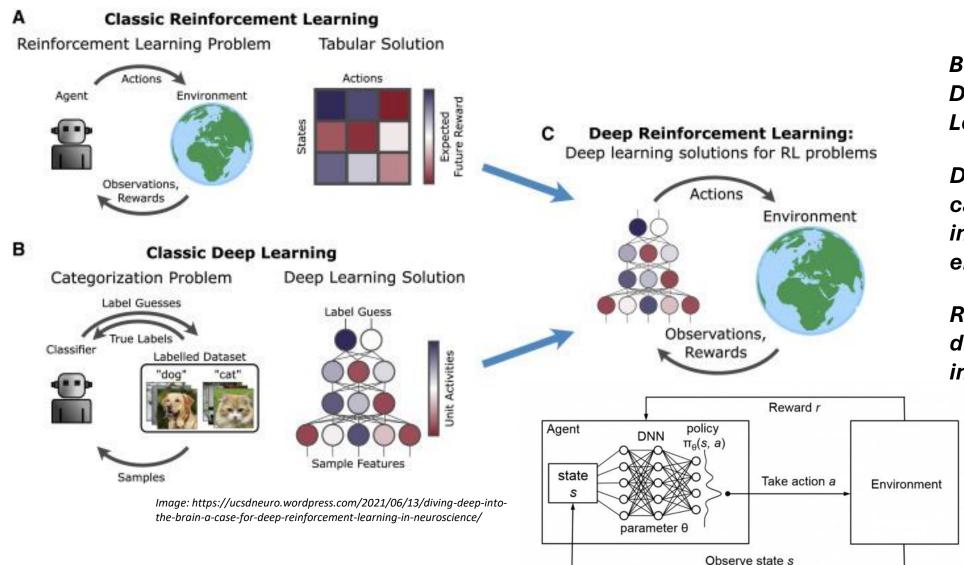
League RoboCup Soccer	2997	799g	1999	2000	, 2007	2007	2003	2004	2005	2006	2007	2008	2009	2010	2017	2012	2013	2014	2015	2016	2017	2012	2019	2020	2023	2022	2023	2024 2026
Simulation 2D				P				7	7	7	7		7	7	7	7	W											
Coach Competition						7	7	7																				
Simulation 3D										P	7		7	7	7	<u> </u>	4	1	4		7	4	7		7	P	4	P
Small Size					7	7	<u> </u>	<u> </u>																				
Middle Size					7	7	7					7	4	7														
Four-Legged																												
Standard Platform																												
Humanoid																												
RoboCupRescue																												
Simulation										7																		
Robot																												
RoboCupJunior																												
Soccer																												
Rescue																												
OnStage																												
Rapidly Manufactured RC																												
RoboCup@Home																												
Open Platform																												
Domestic Stand. Platform																												
Social Standard Platform																												
RoboCupIndustrial																												
RoboCup@Work																												
Logistics																												
Major RoboCup award	₹ Mi	nor a	ward,	/Euro	pean	awa	rd		Offi	cial C	Comp	etitio	n		Exhi	ibitio	n		Part	ticipa	tion			Coll	abora	ation	ation	



Simulation 3D League (Humanoids)



Deep Reinforcement Learning (DRL)

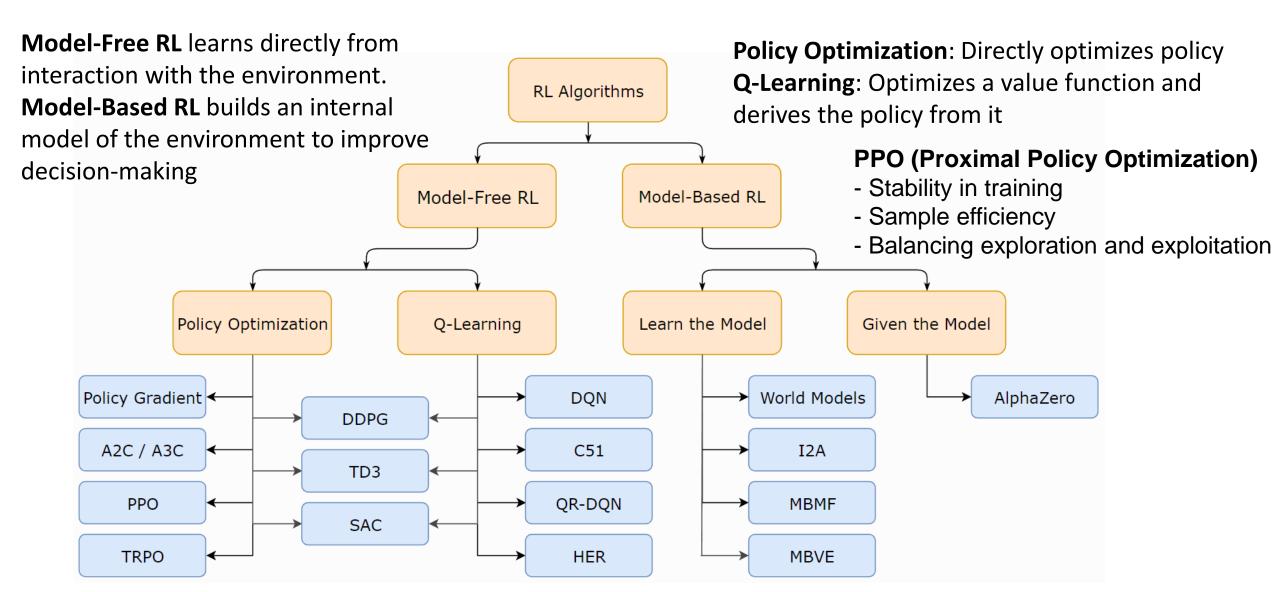


Brains + Experience = Deep Reinforcement Learning

DRL proved machines can outperform humans in dynamic environments

Robots can now learn directly from physical interaction

Deep Reinforcement Learning (DRL)





LIACC

FEUP FACULDADE DE ENGENHARIA

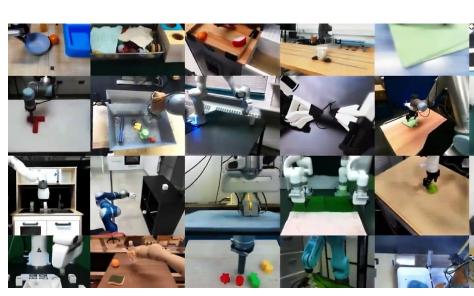
DRL for Learning to Sprint/Dribble





Large Behavior Models - LBMs

- LBMs go beyond processing language. They're designed to interpret, predict,
 - and generate complex sequences of human-like actions and behaviors
- While LLMs focus on text and dialogue, LBMs specialize in behavioural patterns, decision-making processes, and interactive scenarios
- LLMs try to guess the next word, while LBMs try to guess the next action or behaviour
- LBMs are enhancing robots' abilities to move naturally and respond intuitively to their surrounding
- LBMs rely on:
 - Transformer Architecture: Similar to LLMs but adapted to handle sequences of actions/behaviours
 - Multi-Modal Learning: Integrating visual, spatial, and temporal data to enhance understanding
 - Reinforcement Learning: Learning from interactions and feedback to improve behavior over time





Conclusions

- Al has become a Partner no longer a Tool!
- We've travelled from Symbolic Logic to self-learning Agents and Robots
- Impossible to prevent AI use in All Areas of Society and Work
- Agentic AI, LBMs and New GenAI and DRL powered and trained Robots
- Are We Ready for Self-Improving Intelligent Machines?
- Next generation of AI won't just predict it will act, learn, and evolve
- By 2030, Al agents will be integrated into every digital platform/robotic system from industry to healthcare to daily life (LLMs, DRL, LBMs)
- Our challenge is to ensure AI Grows With Us, Not Against Us!









Navigating the Artificial Intelligence Revolution Luís Paulo Reis

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