# [Tech University of Korea Special AI Lecture] The Complete AI Picture - Technology Breakthroughs, Industry Transformation, Market Forces, and Strategic Investment Insights

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# About Speaker

- *Co-Founder & CTO @ Erudio Bio, Inc., San Jose & Novato, CA, USA* 2023 ∼
- *Co-Founder & CEO @ Erudio Bio Korea, Inc., Korea* 2025 ∼
- *Leader of Silicon Valley Privacy-Preserving AI Forum (K-PAI), CA, USA* 2024 ∼
- *Advisor to Korean American Semiconductor Professional Alliance (KASPA)* 2026 ∼
- *CGO / Global Managing Partner @ LULUMEDIC, Seoul, Korea* 2025 ∼
- *KFAS-Salzburg Global Leadership Fellow @ Salzburg Global Seminar, Austria* 2024 ∼
- *Adjunct Professor, EE Department @ Sogang University, Seoul, Korea* 2020 ∼
- *Advisory Professor, EECS Department @ DGIST, Korea* 2020 ∼
- *AI-Korean Medicine Integration Initiative Task Force Member @ The Association of Korean Medicine, Seoul, Korea* 2025 ∼
- *Director of AI Semiconductor @ K-BioX, CA, USA* 2025 ∼
- Global Advisory Board Member @ Innovative Future Brain-Inspired Intelligence System Semiconductor of Sogang University, Korea 2020 ∼
- Technology Consultant @ Gerson Lehrman Gruop (GLG), NY, USA 2022 ∼
- Advisor @ CryptoLab, Inc., Seoul, Korea 2025 ∼

- Co-Founder & CTO / Head of Global R&D / Chief Applied Scientist / Senior Fellow @ Gauss Labs, Inc., Palo Alto, CA, USA 2020 $\sim$ 2023
- Senior Applied Scientist @ Amazon.com, Inc., Vancouver, BC, Canada 2017 $\sim$ 2020
- Principal Engineer @ Software R&D Center, Samsung Electronics 2016 $\sim$ 2017
- Principal Engineer @ Strategic Marketing & Sales, Memory Business 2015 $\sim$ 2016
- Principal Engineer @ DT Team, DRAM Development, Samsung 2012 $\sim$ 2015
- Senior Engineer @ CAE Team, Memory Business, Samsung, Korea 2005 $\sim$ 2012
- PhD - Electrical Engineering @ Stanford University, CA, USA 2001 $\sim$ 2004
- Development Engineer @ Voyan, Santa Clara, CA, USA 2000 $\sim$ 2001
- MS - Electrical Engineering @ Stanford University, CA, USA 1998 $\sim$ 1999
- BS - Electrical & Computer Engineering @ Seoul National University 1994 $\sim$ 1998

# Highlight of Career Journey

- BS in Electrical Engineering (EE) @ Seoul National University
- MS & PhD in Electronics Engineering (EE) @ Stanford University
  - *Convex Optimization - Theory, Algorithms & Software*
  - advisor - *Prof. Stephen P. Boyd*
- Principal Engineer @ Samsung Semiconductor, Inc.
  - *AI & Convex Optimization*
  - collaboration with *DRAM/NAND Design/Manufacturing/Test Teams*
- Senior Applied Scientist @ Amazon.com, Inc.
  - *e-Commerce AIs* - anomaly detection, deep RL, and recommender system
  - *Jeff Bezos's project - drove $200M* in sales via Amazon Mobile Shopping App
- *Co-Founder & CTO / Global R&D Head & Chief Applied Scientist* @ Gauss Labs, Inc.
- *Co-Founder & CTO* @ Erudio Bio, Inc.
- *Co-Founder & CEO* @ Erudio Bio Korea, Inc.

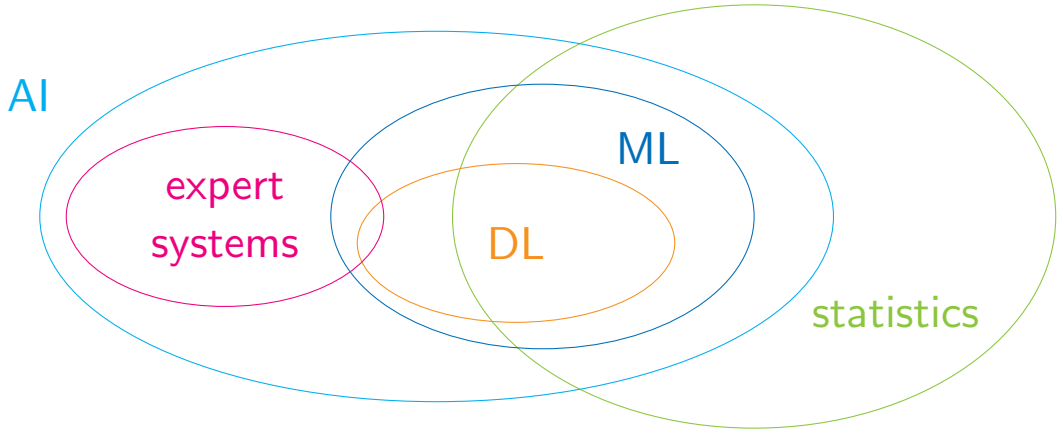
# Unpacking AI-Bio-Silicon Valley Entrepreneurship Landscapes

# Artificial Intelligence

# Definition and History

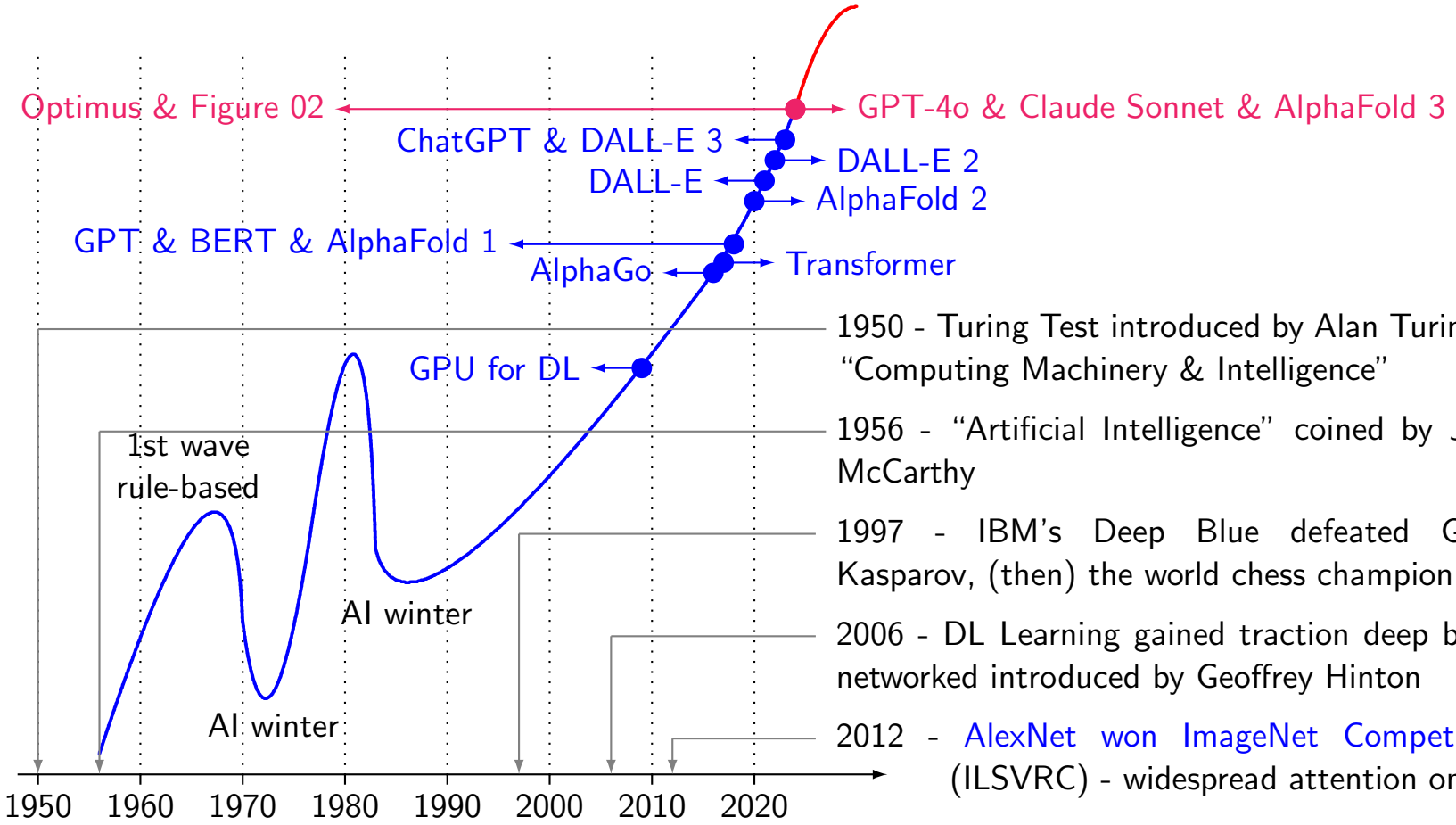
# Definition & relation to other technologies

- AI
  - is technology doing tasks requiring human intelligence, such as learning, problem-solving, decision-making & language understanding
  - encompasses *range of technologies, methodologies, applications & products*
- AI, ML, DL, statistics & expert system[1] [HGH+22]




[1] ML: machine learning & DL: deep learning

# History




1950 - Turing Test introduced by Alan Turing in "Computing Machinery & Intelligence"

1956 - "Artificial Intelligence" coined by John McCarthy

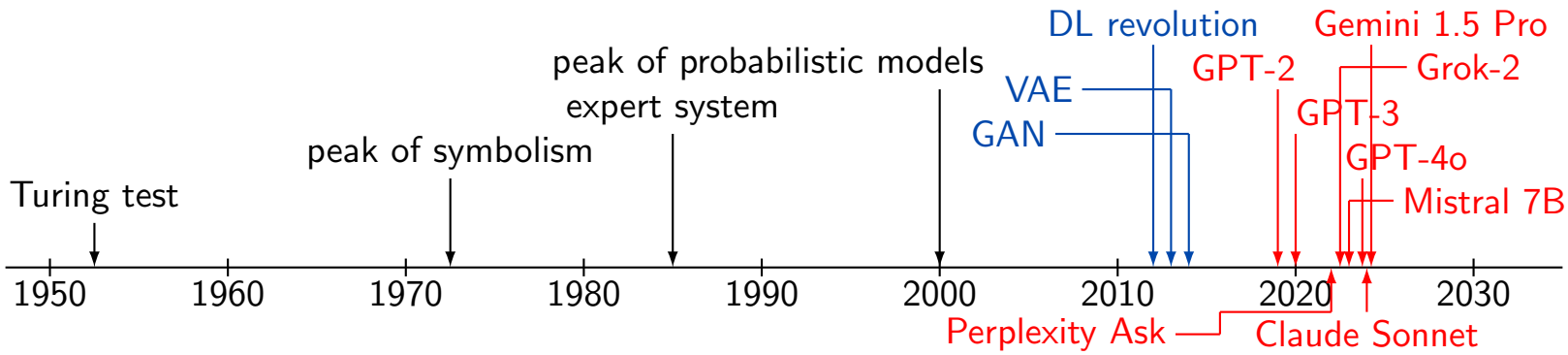
1997 - IBM's Deep Blue defeated Garry Kasparov, (then) the world chess champion

2006 - DL Learning gained traction deep belief networked introduced by Geoffrey Hinton

2012 - AlexNet won ImageNet Competition (ILSVRC) - widespread attention on DL

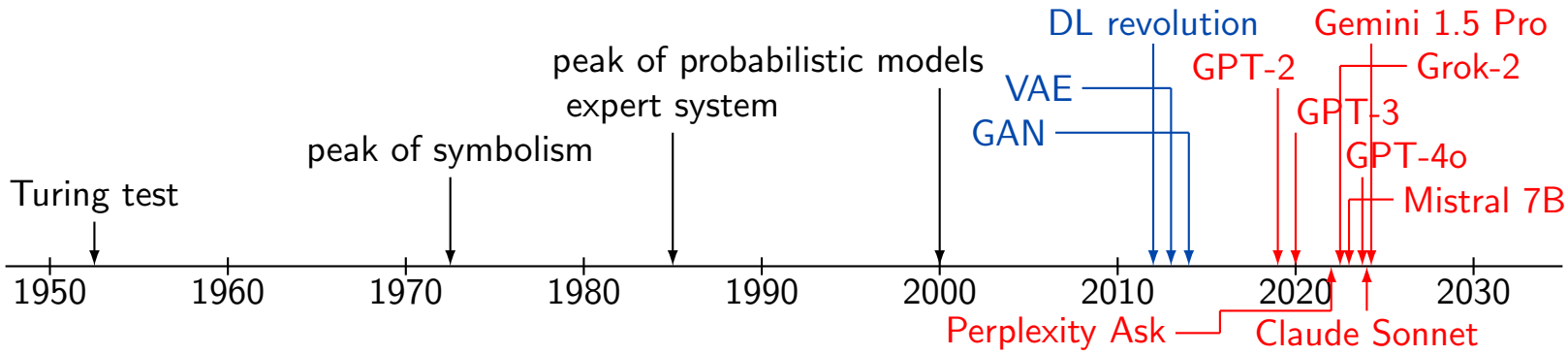
# Birth of AI - early foundations & precursor technologies

- 1950s $\sim$ 1970s
  - Alan Turing - concept of *"thinking machine" & Turing test* to evaluate machine intelligence (1950s)
  - *symbolists* (as opposed to connectionists) - early AI focused on symbolic reasoning, logic & problem-solving - Dartmouth Conference in 1956 by *John McCarthy, Marvin Minsky, Allen Newell & Herbert A. Simon*
  - precursor technologies - genetic algorithms (GAs), Markov chains & *hidden Markov models (HMMs)* - laying foundation for generative processes (1970s $\sim$)



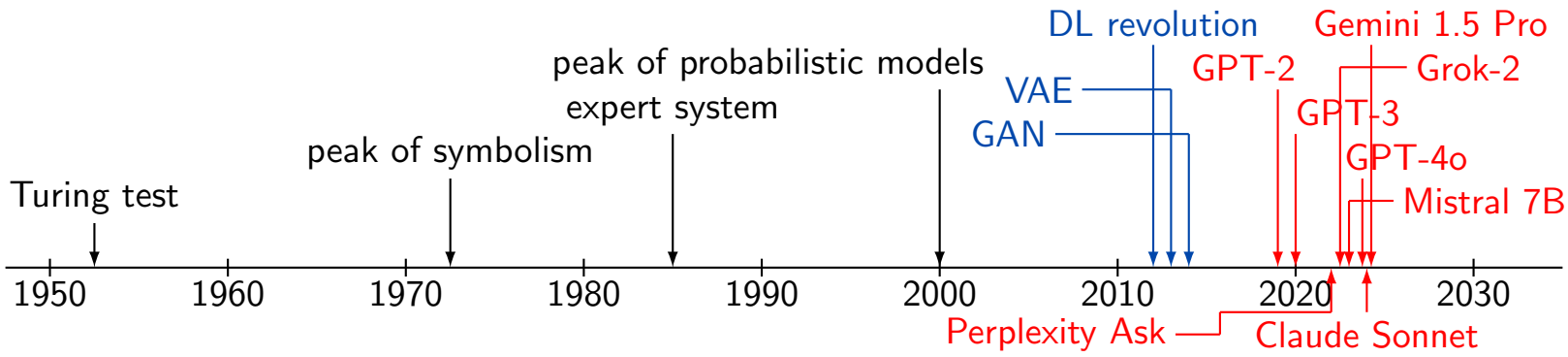
# Rule-based systems & probabilistic models

- 1980s $\sim$ early 2000s
  - *expert systems* (1980s) - AI systems designed to mimic human decision-making in specific domains
  - development of neural networks (NN) w/ backpropagation *training multi-layered networks* - setting stage for way more complex generative models
  - *probabilistic models* (including network models, $i.e.$, Bayesian networks) & Markov models - laying groundwork for data generation & pattern prediction



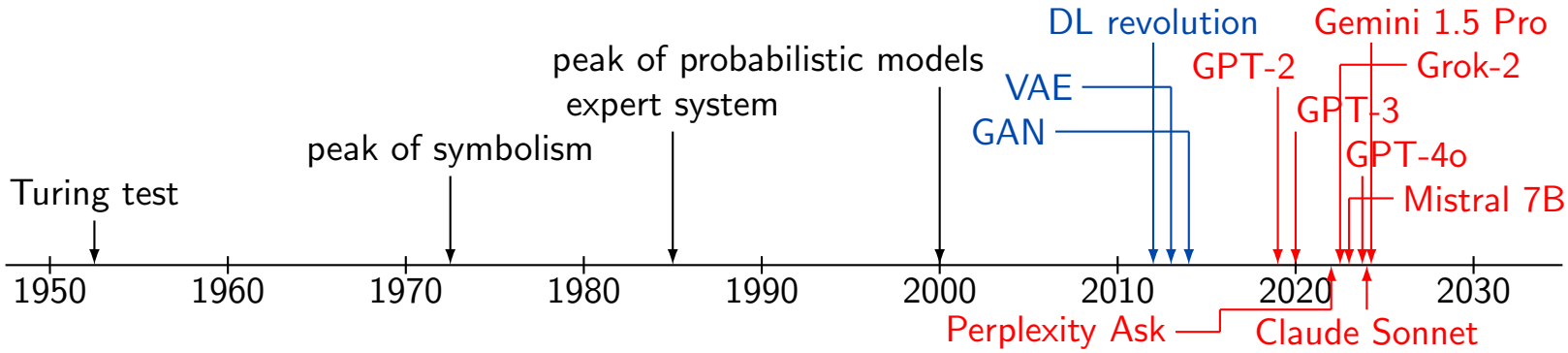
# Rise of deep learning & generative models

- 2010s - breakthrough in genAI
  - *deep learning (DL) revolution* - advances in GPU computing and data availability led to the rapid development of deep neural networks.
  - *variational autoencoder (VAE)* (2013) - by Kingma and Welling - learns mappings between input and latent spaces
  - *generative adversarial network (GAN)* (2014) - by Ian Goodfellow - game-changer in generative modeling where two NNs compete each other to create realistic data
    - widely used in image generation & creative tasks



# Transformer models & multimodal AI

- late 2010s $\sim$ Present
  - Transformer architecture (2017) - by Vaswani et al.
    - *revolutionized NLP, $e.g.$*, LLM & various genAI models
  - GPT series - generative pre-trained transformer
    - GPT-2 (2019) - generating human-like texts - *marking leap in language models*
    - GPT-3 (2020) - 175B params - set *new standards for LLM*
  - multimodal systems - DALL-E & CLIP (2021) - *linking text and visual data*
  - emergence of diffusion models (2020s) - new approach for generating high-quality images - progressively "denoising" random noise (DALL-E 2 & Stable Diffusion)



# Significant AI Achievements - 2014 – 2025

# Deep learning revolution

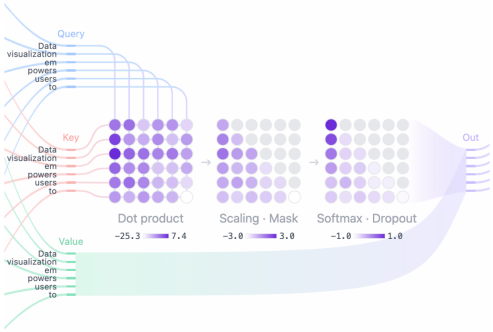
- 2012 – 2015 - DL revolution[2]
  - CNNs demonstrated exceptional performance in image recognition, $e.g.$, *AlexNet's victory in ImageNet competition*
  - widespread adoption of DL learning in CV transforming industries
- 2016 - AlphaGo defeats human Go champion
  - DeepMind's AlphaGo defeated world champion in Go, extremely complex game *believed to be beyond AI's reach*
  - significant milestone in RL - AI's potential in solving complex & strategic problems



[2] CV: computer vision, NN: neural network, CNN: convolutional NN, RL: reinforcement learning

# Transformer changes everything

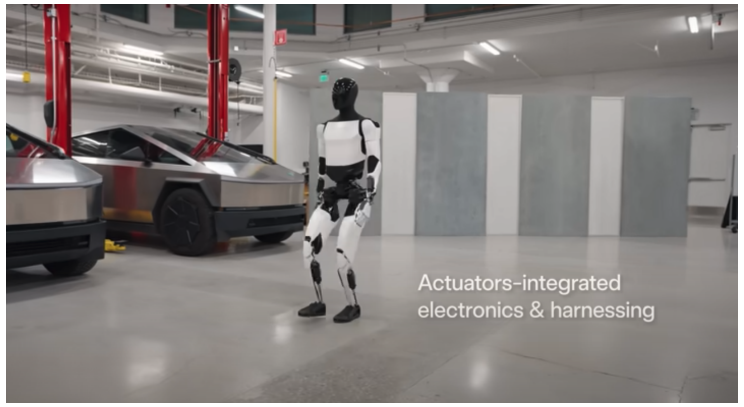
- 2017 – 2018 - Transformers & NLP breakthroughs[3]
  - *Transformer (e.g., BERT & GPT) revolutionized NLP*
  - major advancements in, *e.g.*, machine translation & chatbots
- 2020 - AI in healthcare – AlphaFold & beyond
  - DeepMind's *AlphaFold solves 50-year-old protein folding problem* predicting 3D protein structures with remarkable accuracy
  - accelerates drug discovery and personalized medicine - offering new insights into diseases and potential treatments


[3] NLP: natural language processing, GPT: generative pre-trained transformer

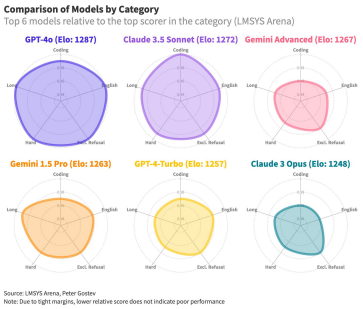
# Lots of breakthroughs in AI technology and applications in 2024

- proliferation of advanced AI models
  - GPT-4o, Claude Sonnet, Claude 3 series, Llama 3, Sora, Gemini
  - *transforming industries* such as content creation, customer service, education, $etc.$
- breakthroughs in specialized AI applications
  - Figure 02, Optimus, AlphaFold 3
  - driving unprecedented advancements in automation, drug discovery, scientific understanding - *profoundly affecting healthcare, manufacturing, scientific research*



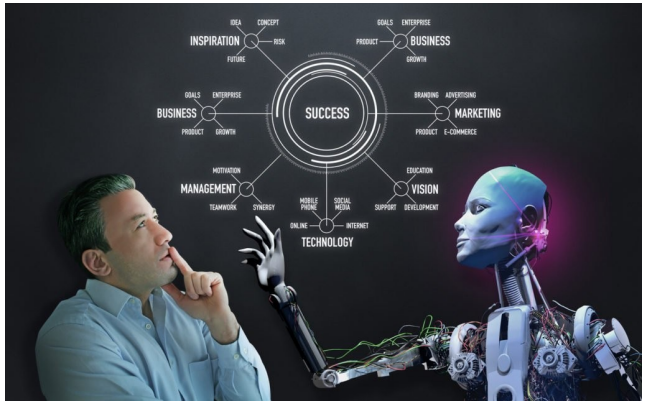
# Major AI Breakthroughs in 2025

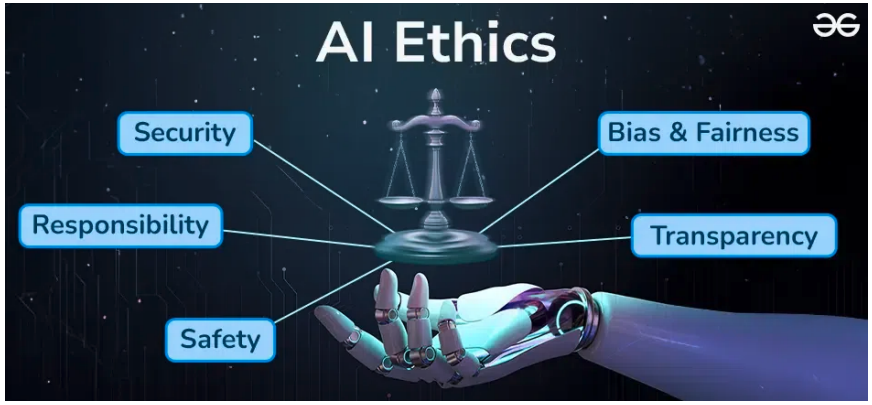
- next-generation foundation models
  - GPT-5 and Claude 4 demonstrate emergent reasoning abilities
  - open-source models achieving parity with leading commercial systems from 2024
- hardware innovations
  - NVIDIA's Blackwell successor architecture delivering 3-4x performance improvement
  - AMD's MI350 accelerators challenging NVIDIA's market dominance
- AI-human collaboration systems
  - seamless multimodal interfaces enabling natural human-AI collaboration
  - AI systems effectively explaining reasoning and recommendations
  - augmented reality interfaces providing real-time AI assistance in professional contexts

# Transformative impact of AI - reshaping industries, work & society

- accelerating human-AI collaboration
  - not only reshaping industries but *altering how humans interact with technology*
  - AI's role as collaborator and augmentor redefines productivity, creativity, the way we address global challenges, $e.g.$, *sustainability & healthcare*
- AI-driven automation *transforms workforce dynamics* - creating new opportunities while challenging traditional job roles
- *ethical AI considerations* becoming central not only to business strategy, but to society as a whole - *influencing regulations, corporate responsibility & public trust*

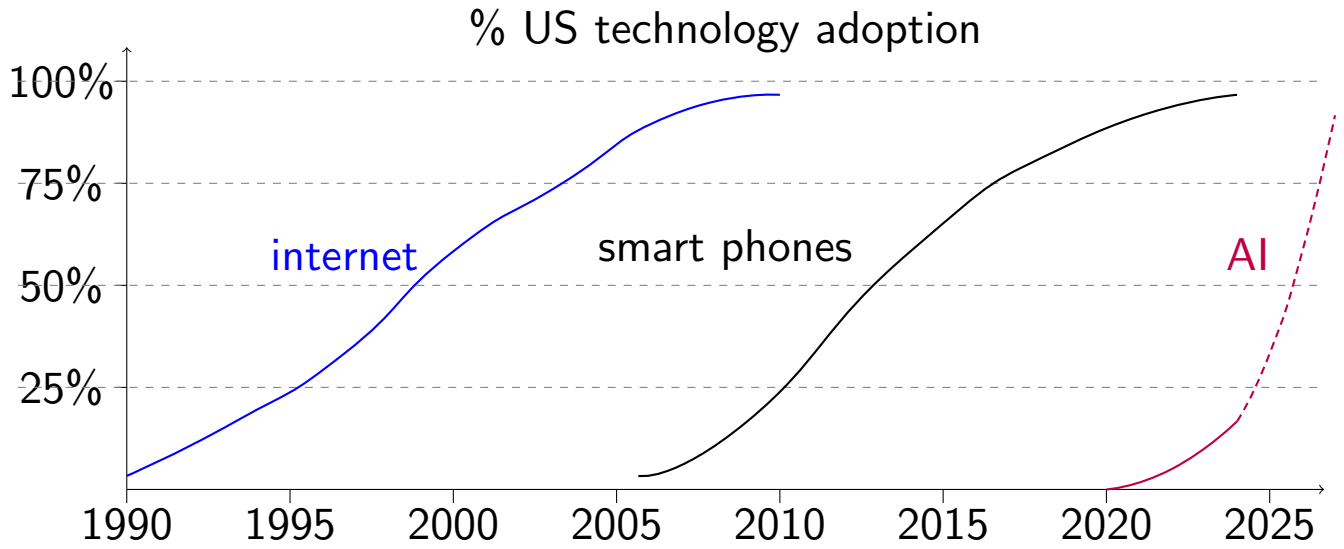


# Measuring AI's Ascent

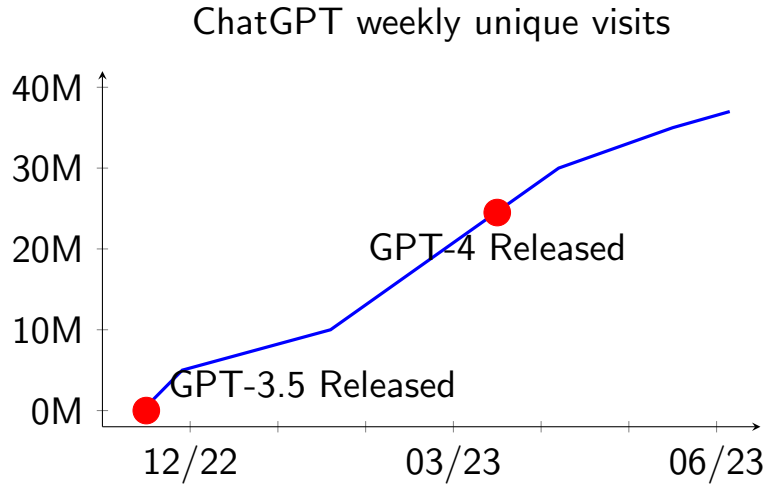
# Where are we in AI today?

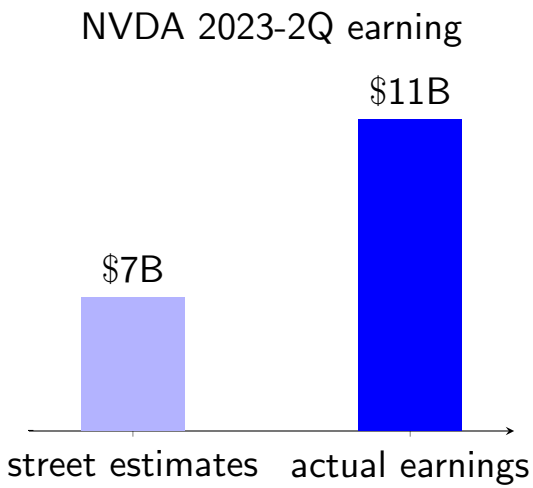
- sunrise phase - currently experiencing dawn of AI era with significant advancements and increasing adoption across various industries
- early adoption - in early stages of AI lifecycle with widespread adoption and innovation across sectors marking significant shift in technology's role in society



# Explosion of AI ecosystems - ChatGPT & NVIDIA

- took only *5 months for ChatGPT users to reach 35M*
- NVDIA 2023 Q2 earning exceeds market expectation by big margin - $7B vs $13.5B
  - surprisingly, *101% year-to-year growth*
  - even more surprisingly *gross margin was 71.2%* - up from 43.5% in previous year[4]



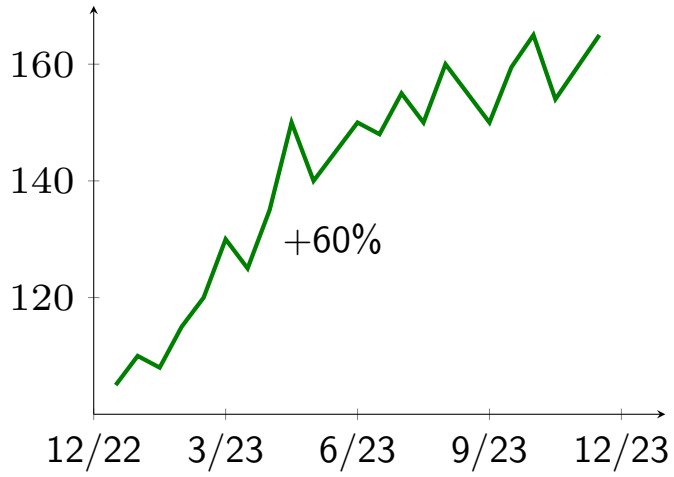




[4] source - Bloomberg

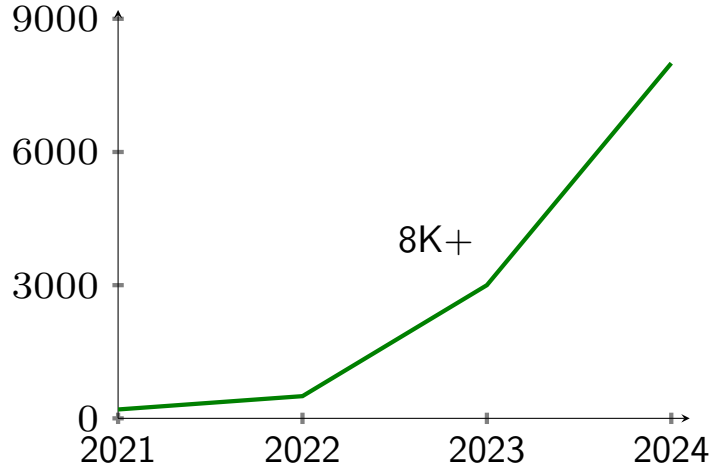
# Explosion of AI ecosystems - AI stock market

- *AI investment surge in 2023 - portfolio performance soars by 60%*
  - AI-focused stocks significantly outpaced traditional market indices
- *over 8,000 new AI applications* developed in last 3 years
  - applications span from healthcare and finance to manufacturing and entertainment

AI basket performance - 2023

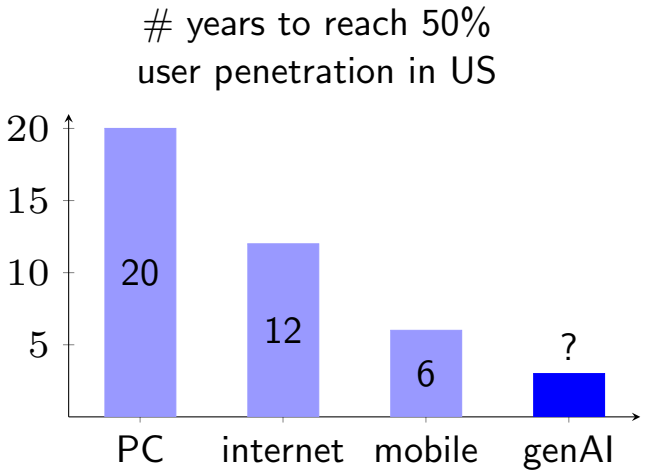


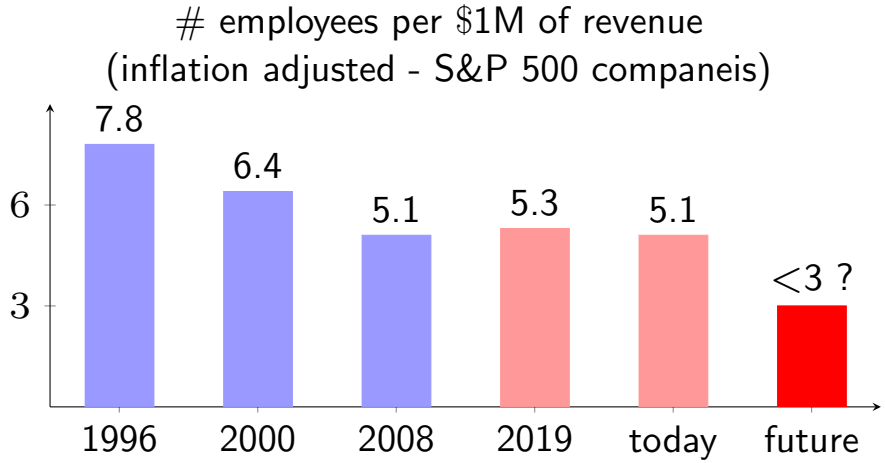

new AI applications over time



# AI's transformative impact - adoption speed & economic potential

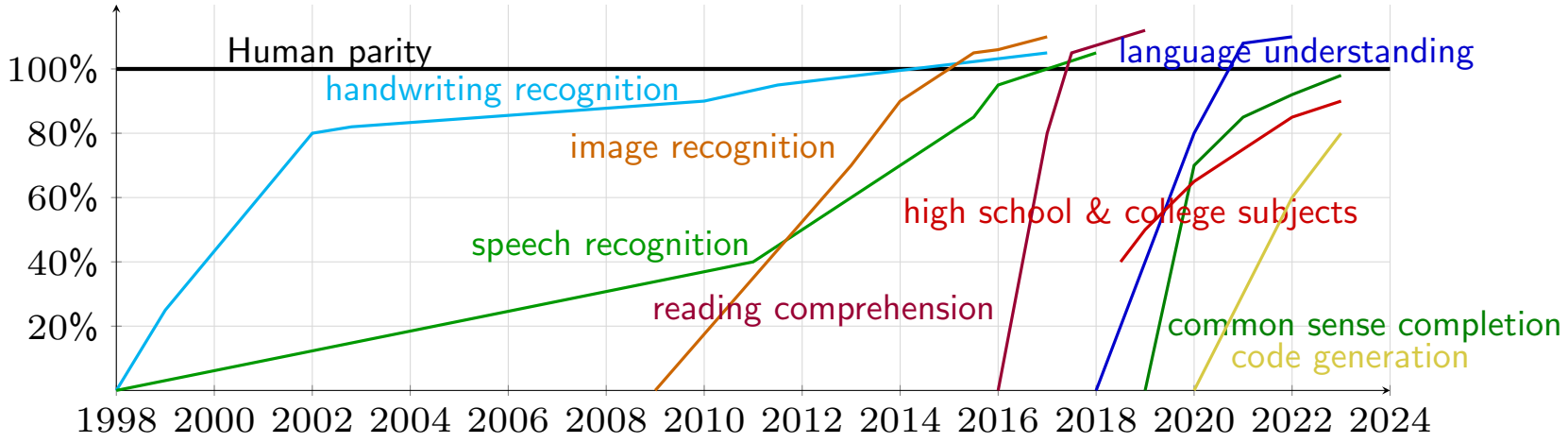
- adoption - has been twice as fast with platform shifts suggesting
  - increasing demand and readiness for new technology improved user experience & accessibility
- AI's potential to drive economy for years to come
  - 35% improvement in productivity driven by introduction of PCs and internet
  - greater gains expected with AI proliferation



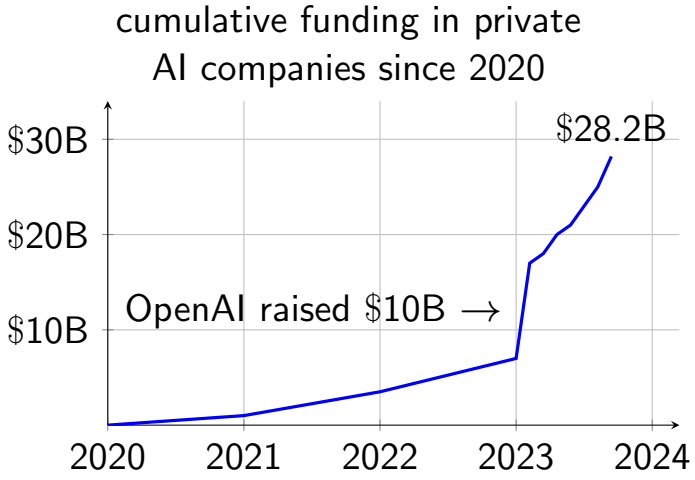
# AI getting more & more faster

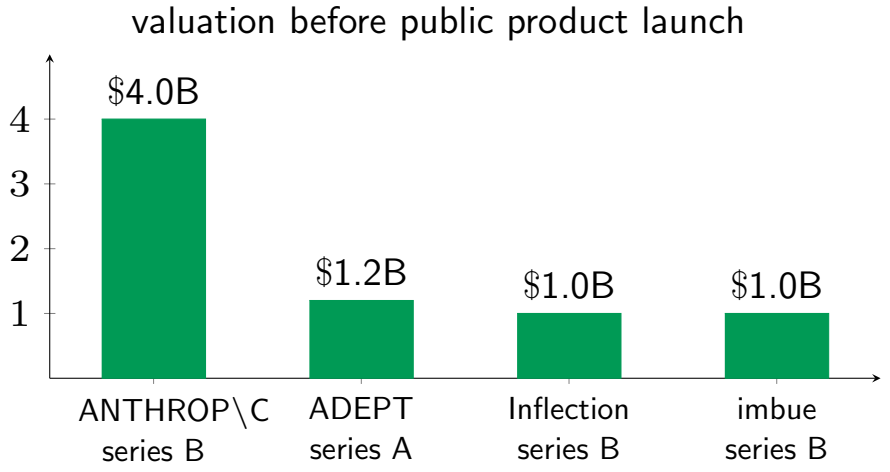
- steep upward slopes of AI capabilities highlight accelerating pace of AI development
  - period of exponential growth with AI potentially mastering new skills and surpassing human capabilities at ever-increasing rate
- closing gap to human parity - some capabilities approaching or arguably reached human parity, while others having still way to go
  - achieving truly human-like capabilities in broad range remains a challenge



# Massive investment in AI

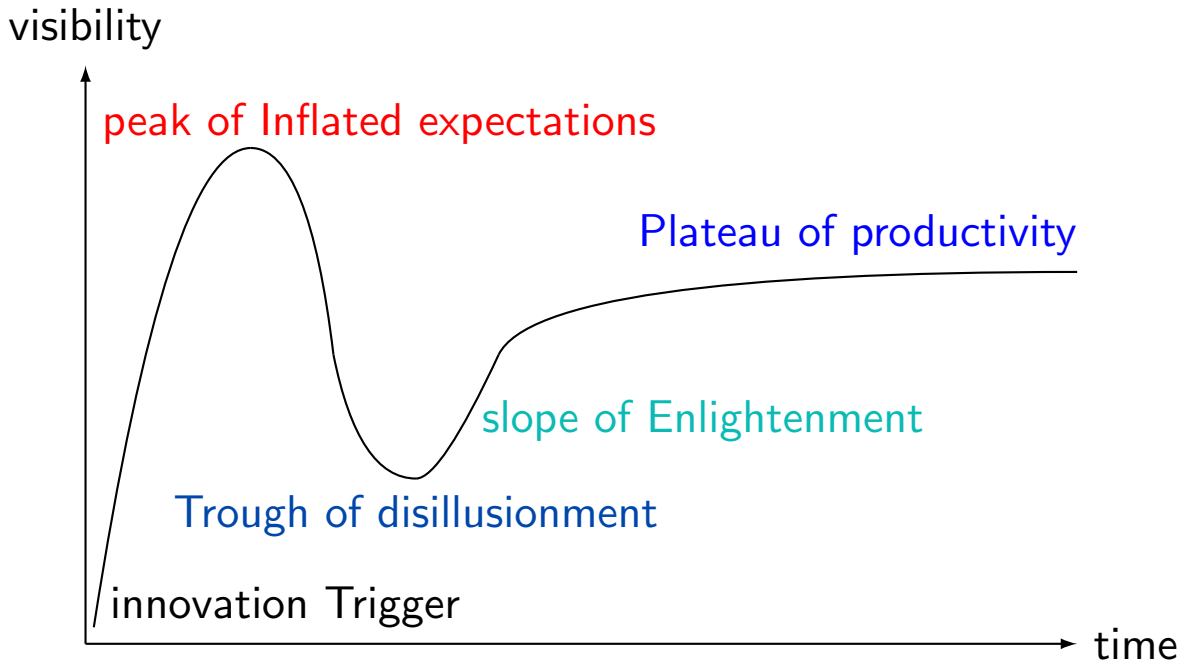
- *explosive growth* - cumulative funding skyrocketed reaching staggering $28.2B
- OpenAI - significant fundraising (=$10B) fueled rapid growth
- *valuation surge* - substantial valuations even before public products for stella companies
- *fierce competition for capital* among AI startups driving innovation & accelerating development
- massive investment indicates *strong belief in & optimistic outlook for potential of AI* to revolutionize industries & drive economic growth



# Is AI hype?

# Technology hype cycle




- innovation trigger - technology breakthrough kicks things off
- peak of inflated expectations - early publicity induces many successes followed by even more
- trough of disillusionment - expectations wane as technology producers shake out or fail
- slope of enlightenment - benefit enterprise, technology better understood, more enterprises fund pilots

# Fiber vs cloud infrastructure

- fiber infrastructure - 1990s
  - Telco Co's raised $1.6T of equity & $600B of debt
  - bandwidth costs decreased 90% within 4 years
  - companies - Covage, NothStart, Telligent, Electric Lightwave, 360 networks, Nextlink, Broadwind, UUNET, NFS Communications, Global Crossing, Level 3 Communications
  - became *public good*
- cloud infrastructure - 2010s
  - entirely new computing paradigm
  - mostly public companeis with data centers
  - *big 4 hyperscalers generate* $150B + annual revenue

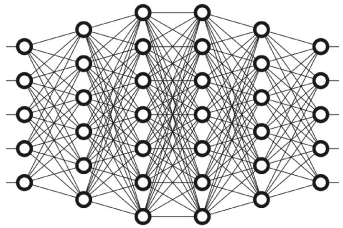
# Yes & No

| characteristics of hype cycles | speaker's views |
|---|---|
| value accrual misaligned with investment | • OpenAI still operating at a loss; business model *still* not clear<br>• gradual value creation across broad range of industries and technologies ($e.g.$, CV, LLMs, RL) unlike fiber optic bubble in 1990s |
| overestimating timeline & capabilities of technology | • self-driving cars delayed for over 15 years, with limited hope for achieving level 5 autonomy<br>• AI, however, has proven useful within a shorter 5-year span, with enterprises eagerly adopting |
| lack of widespread utility due to technology maturity | • AI already providing significant utility across various domains<br>• vs quantum computing remains promising in theory but lacks widespread practical utility |

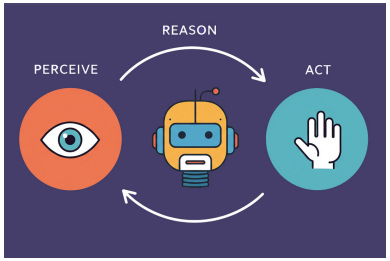
# AI Agents

# AI progress in 21st century in keywords

- 2010 $\sim$ Big Data
- 2012 $\sim$ Deep Learning
- 2017 $\sim$ Transformer - Attention is All you need!
- 2022 $\sim$ LLM & genAI
- 2024 $\sim$ AI Agent (Agentic AI)

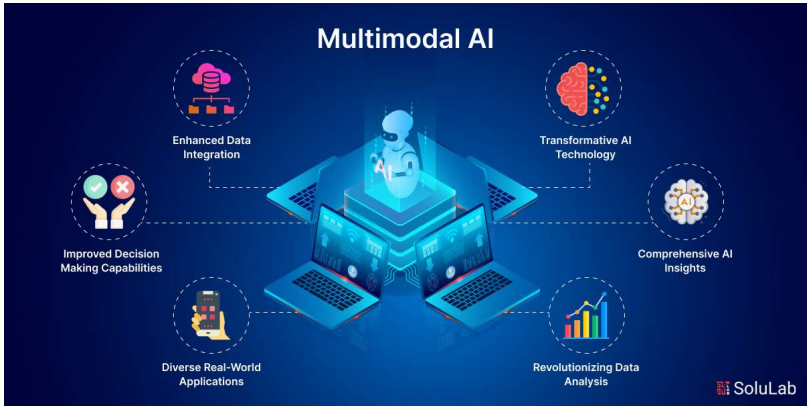



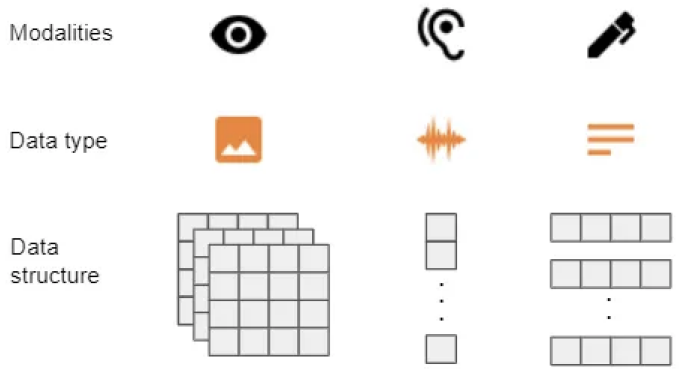




# Multimodal learning

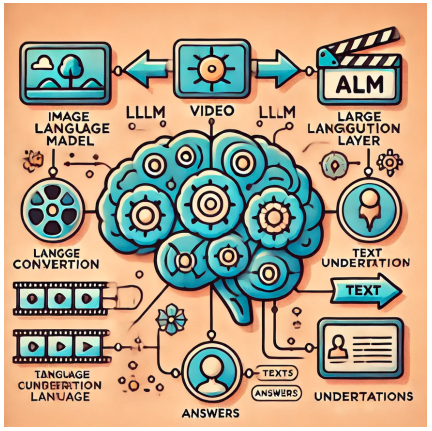
- understand information from multiple modalities, $e.g.$, text, images, audio, video
- representation learning methods
  - combine multiple representations or learn multimodal representations simultaneously
- applications
  - images from text prompt, videos with narration, musics with lyrics
- collaboration among different modalities
  - understand image world (open system) using language (closed system)



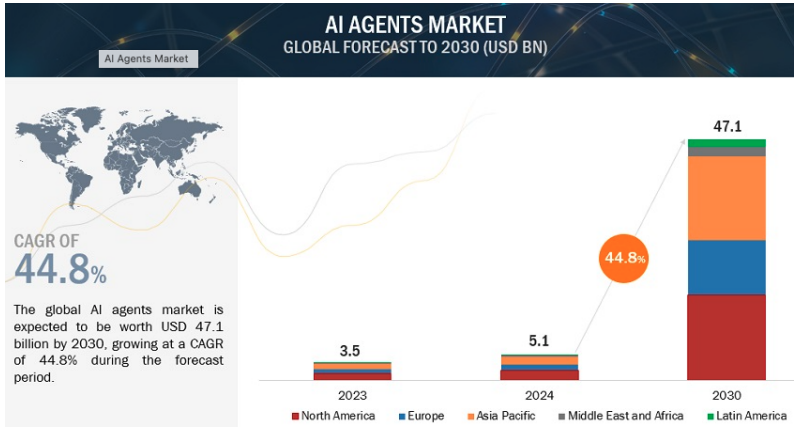
# Implications of success of LLMs

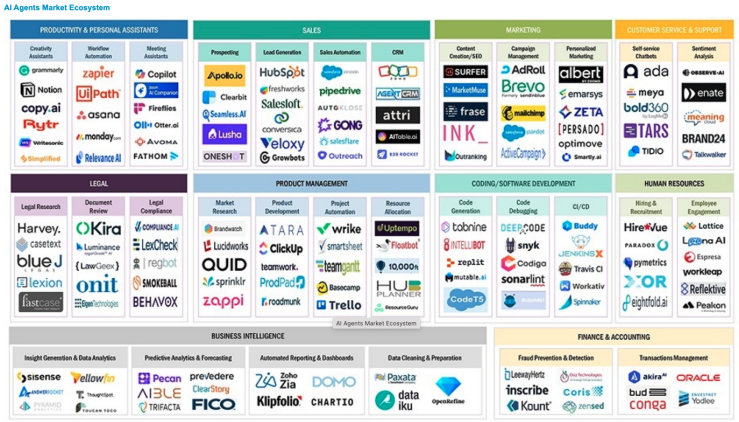
- many researchers change gears towards LLM
  - from computer vision (CV), speach, music, video, even reinforcement learning
- *LLM is not only about NLP . . .* humans have . . .
  - evolved to optimize natural language structures for eons
  - handed down knowledge using *this natural languages* for thousands of years
  - internal structure (or equivalently, representation) of natural languages optimized via *thousands of generation by evolution*
- LLM *connects non-linguistic world (open system) via natural languages (closed system)*




# Multimodal AI (mmAI)

- mmAI - systems processing & integrating data from multiple sources & modalities, to generate unified response / decision
- 1990s – 2000s - early systems - initial research combining basic text & image data
- 2010s - CNNs & RNNs enabling more sophisticated handling of multimodality
- 2020s - modern multimodal models - Transformer-based architectures handling complex multi-source data at highly advanced level
- mmAI *mimics human cognitive ability* to interpret and integrate information from various sources, leading to holistic decision-making



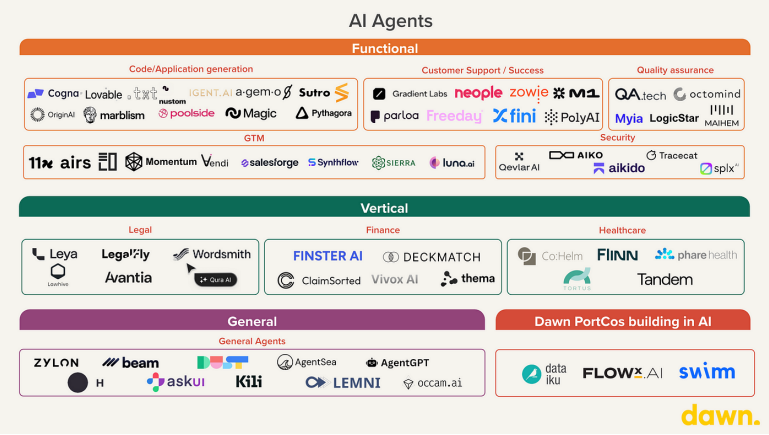



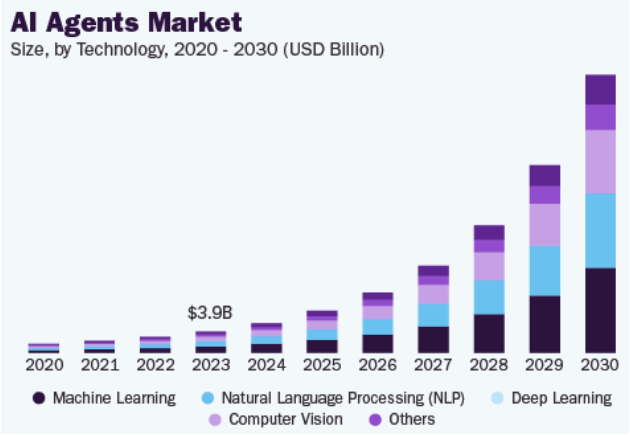
# mmAI Technology

- core components
  - data preprocessing - images, text, audio & video
  - architectures - unified Transformer-based ($e.g.$, ViT) & cross-attention mechanisms / hybrid architectures ($e.g.$, CNNs + LLMs)
  - integration layers - fusion methods for combining data representations from different modalities
- technical challenges
  - data alignment - accurate alignment of multimodal data
  - computational demand - high-resource requirements for training and inferencing
  - diverse data quality - manage variations in data quality across modalities
- advancements
  - multimodal embeddings - shared feature spaces interaction between modalities
  - self-supervised learning - leverage unlabeled data to learn representations across modalities

# AI agents powered by multimodal LLMs

- foundation
  - integrate multimodal AI capabilities for enhanced interaction & decision-making
- components
  - perceive environment through multiple modalities (visual, audio, text), process using LLM technology, generate contextual responses & take actions
- capabilities
  - understand complex environments, reason across modalities, engage in natural interactions, adapt behavior based on context & feedback



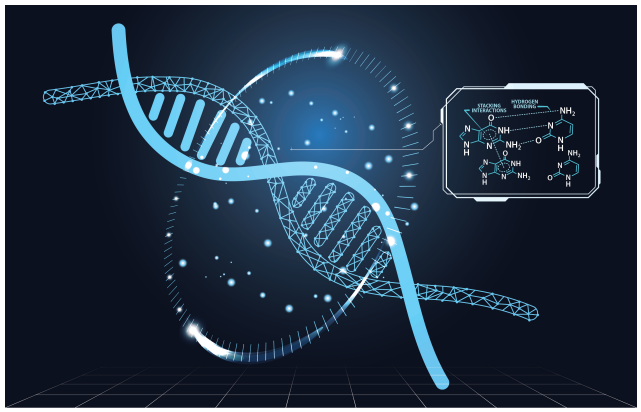
# AI agents - Present & Future

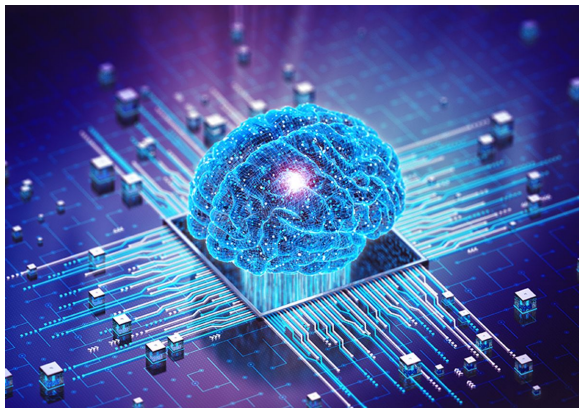
- emerging applications
  - scientific research - agents analyzing & running experiments & generating hypotheses
  - creative collaboration - AI partners in design & art combining multiple mediums
  - environmental monitoring - processing satellite sensor data for climate analysis
  - healthcare - enhanced diagnostic combining imaging, $e.g.$, MRI, with patient history
  - customer experience - virtual assistants understanding spoken language & visual cues
  - autonomous vehicles - integration of visual, radar & audio data
- future
  - ubiquitous AI agents - seamless integration into everyday devices
  - highly tailored personalized experience - in education, entertainment & healthcare

# AI & Biotech

# AI in biology

- AI has been used in biological sciences, and science in general
- AI's ability to process large amounts of raw, unstructured data ($e.g.$, DNA sequence data)
  - reduces time and cost to conduct experiments in biology
  - enables others types of experiments that previously were unattainable
  - contributes to broader field of engineering biology or biotechnology
- AI increases human ability to make direct changes at cellular level and create novel genetic material ($e.g.$, DNA and RNA) to obtain specific functions





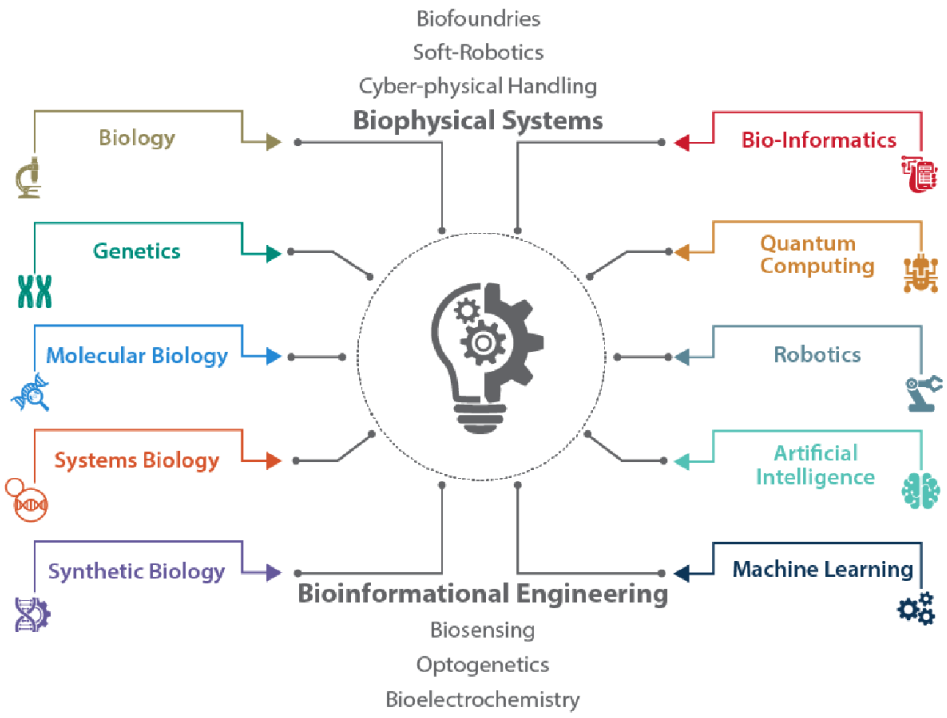
# Biotech

# Biotech

- biotechnology
  - is multidisciplinary field leveraging broad set of sciences and technologies
  - relies on and builds upon advances in other fields such as nanotechnology & robotics, and, increasingly, AI
  - enables researchers to read and write DNA
    - sequencing technologies "read" DNA while gene synthesis technologies take sequence data and "write" DNA turning data into physical material
- 2018 National Defense Strategy & Senior US Defense and Intelligence Officials identified emerging technologies that could have disruptive impact on US national security [Say21]
  - *AI*, lethal autonomous weapons, hypersonic weapons, directed energy weapons, *biotechnology,* quantum technology
- other names for biotechnology are engineering biology, synthetic biology, biological science (when discussed in context of AI)

# Biotech - multidisciplinary field

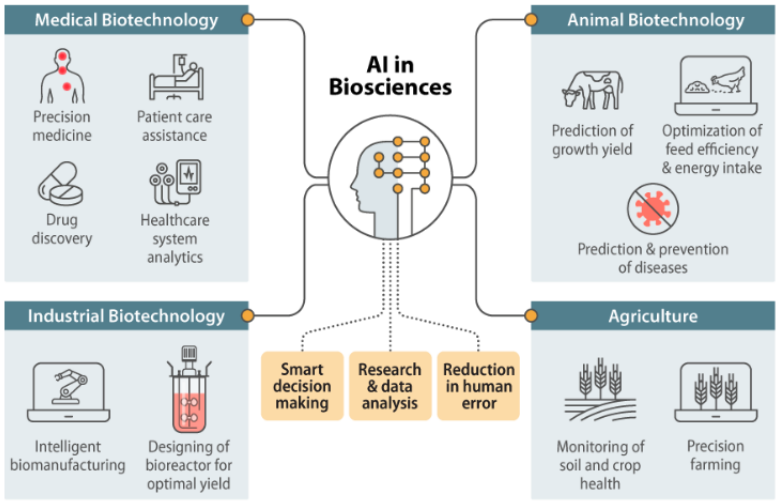
- sciences and technologies enabling biotechnology include (but not limited to)
  - (molecular) biology, genetics, systems biology, synthetic biology, bio-informatics, quantum computing, robotics [DFJ22]



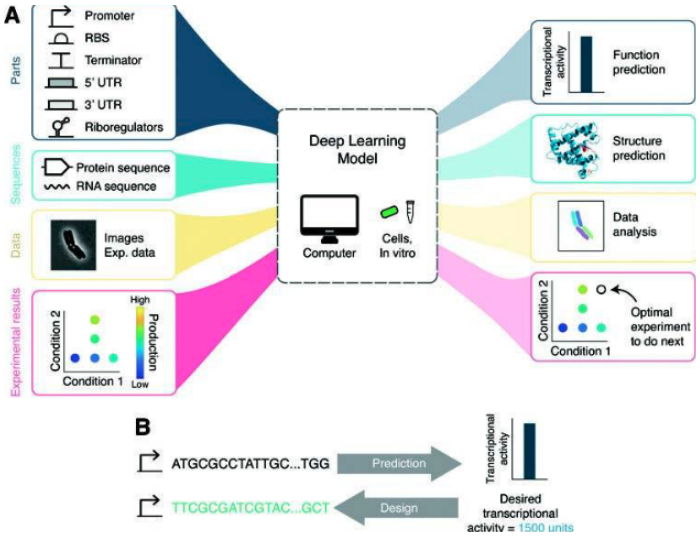
# Convergence of AI and biological design

- AI & biological sciences converging [BKP22]
  - each building upon the other's capabilities for new research and development across multiple areas
- Demis Hassabis, CEO & cofounder of DeepMind, said of biology [Toe23]

  ". . . biology can be thought of as information processing system, albeit extraordinarily complex and dynamic one . . . just as mathematics turned out to be the right description language for physics, biology may turn out to be *the perfect type of regime for the application of AI!*"
- both AI & biotech rely on and build upon advances in other scientific disciplines and technology fields, such as nanotechnology, robotics, and increasingly big data ($e.g.$, genetic sequence data)
  - each of these fields itself convergence of multiple sciences and technologies
- so *their impacts can combine to create new capabilities*



# Multi-source genetic sequence data




- AI, essential to analyzing exponential growth of genetic sequence data

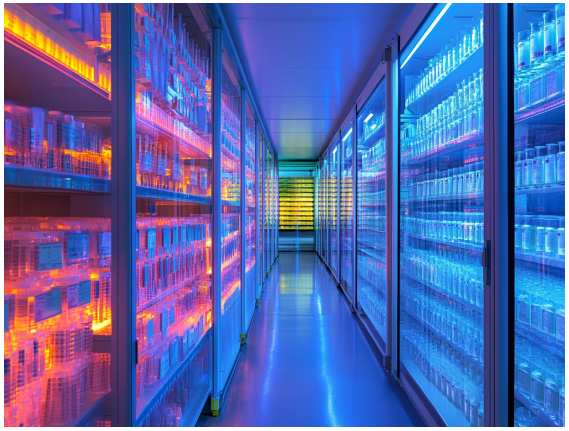
  "AI will be essential to fully understanding how genetic code interacts with biological processes" - US National Security Commission on Artificial Intelligence (NSCAI)

  – process huge amounts of biological data, $e.g.$, genetic sequence data, coming from different biological sources for understanding complex biological systems
    - sequence data, molecular structure data, image data, time-series, omics data

- $e.g.$, analyze genomic data sets to determine the genetic basis of particular trait and potentially uncover genetic markers linked with that trait

# Quality & quantity of biological data

- limiting factor, however, is *quality and quantity* of biological data, $e.g.$, DNA sequences, that AI is trained on
  - $e.g.$, accurate identification of particular species based on DNA requires reference sequences of *sufficient quality* to exist and be available
- databases have varying standards - access, type, and quality of information
- design, management, quality standards, and data protocols for reference databases can affect utility of particular DNA sequence



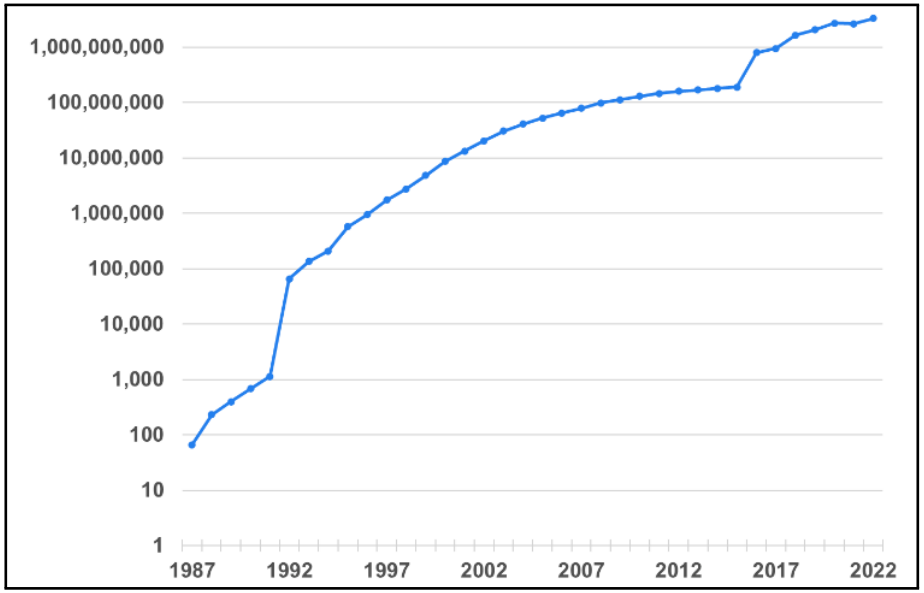
# Rapid growth of biological data

- volume of genetic sequence data grown exponentially as sequencing technology evolved
- more than 1,700 databases incorporating data on genomics, protein sequences, protein structures, plants, metabolic pathways, $etc.$, $e.g.$
  - open-source public database
    - Protein Data Bank, US-funded data center - more than *terabyte of three-dimensional structure data* for biological molecules, $e.g.$, proteins, DNA, RNA
  - proprietary database
    - Gingko Bioworks - more than *2B protein sequences*
  - public research groups
    - Broad Institute - produces roughly *500 terabases of genomic data per month*
- great potential value in aggregate volume of genetic datasets that can be collectively mined to discover and characterize relationships among genes

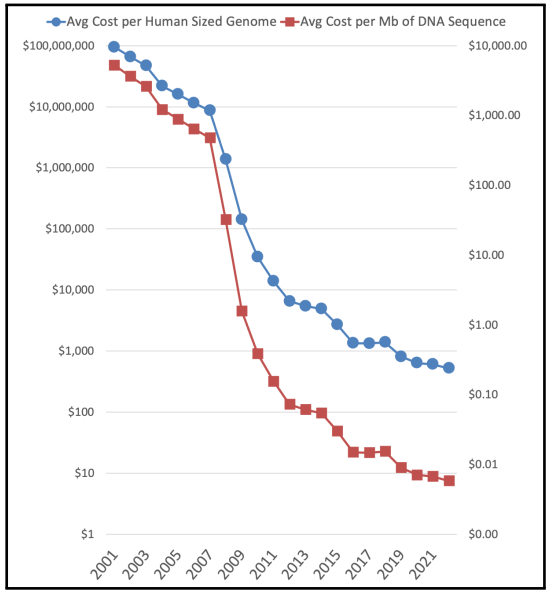
# Volume and sequencing cost of DNA over time

- volume of DNA sequences & DNA sequencing cost
  - data source: National Human Genome Research Institute (NHGRI) [Wet23] & International Nucleotide Sequence Database Collaboration (INSDC)
- *more dramatic than Moore's law!*

# sequences in INSDC




DNA sequencing cost



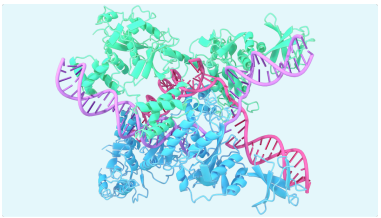
# Bio data availability and bias

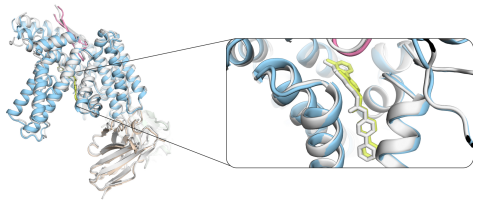
- US National Security Commission on Artificial Intelligence (NSCAI) recommends
  - US fund and prioritize development of a biobank containing *"wide range of high-quality biological and genetic data sets securely accessible by researchers"*
  - establishment of database of broad range of human, animal, and plant genomes would
    - *enhance and democratize biotechnology innovations*
    - *facilitate new levels of AI-enabled analysis of genetic data*
- bias - availability of genetic data & decisions about selection of genetic data can introduce bias, $e.g.$
  - training AI model on datasets emphasizing or omitting certain genetic traits can affect how information is used and types of applications developed - *potentially privileging or disadvantaging certain populations*
  - access to data and to AI models themselves may impact communities of differing socioeconomic status or other factors unequally

# Emerging Trends in Biotech

# AlphaFold

- solving 50-year-old protein folding problem, *"one of biology's grand challenges"*
  - definition - given amino acid sequence, predict how it folds into a 3D structure
  - proteins fold in microseconds, but predicting computationally nearly impossible
- AlphaFold 1 (2018) - DL + physics-based energy functions → AlphaFold 2 (2020) - attention-based NN solving protein folding "in principle" → AlphaFold 3 (2024) - diffusion-based DL, drug-protein interactions, protein complexes
- AlphaFold protein structure database
  - >200MM protein structures - nearly every known protein, used by >2MM researchers
- Applications & implications
  - drug discovery - target identification, lead optimization, side effect prediction
  - enzyme engineering, agriculture, environmental, vaccine development



# AlphaGo

- Go board marked with 19×19 grid - $(19 \times 19)! = 361! \approx 1.44 \times 10^{768}$
  1437923258884890654832362511499863354754907538644755876127282765299227795534389618856841908003141196
  0714137944348905859683839682333043216077138088370565578796691924861827097800358990211005794501073330
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  9917299937133680728459000034496420337066440853337001284286412654394495050773954560000000000000000000
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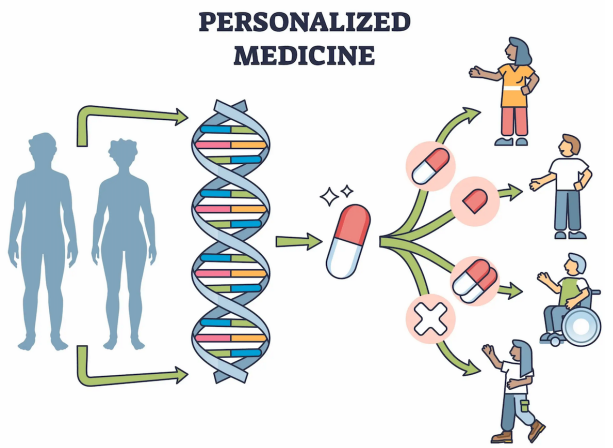
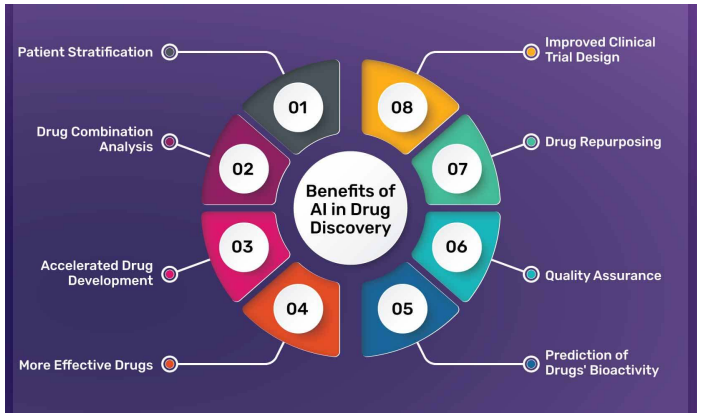
- deep reinforcement learning with Monte Carlo tree search
  - trained on thousands of years of Go game history
  - AlphaGo Zero learns by playing against itself
- development experience, insight, knowledge, know-how transferred to AlphaFold

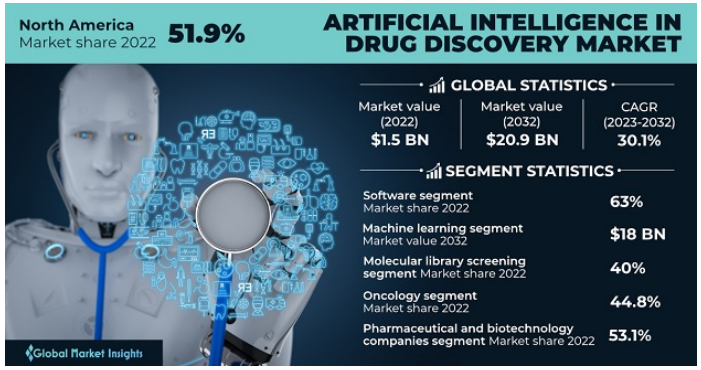

# Personalized medicine

- *shift from one-size-fits-all approach to tailored treatments*
- based on individual genetic profiles, lifestyles & environments
- AI enables analysis of vast data to predict patient responses to treatments, thus enhancing efficacy and reducing adverse effects
- $e.g.$
  - custom cancer therapies
  - personalized treatment plans for rare diseases
  - precision pharmacogenomics
- companies - Tempus, Foundation Medicine, $etc.$



# AI-driven drug discovery

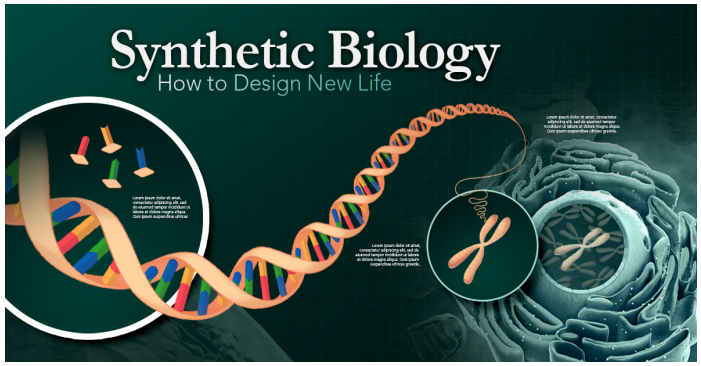


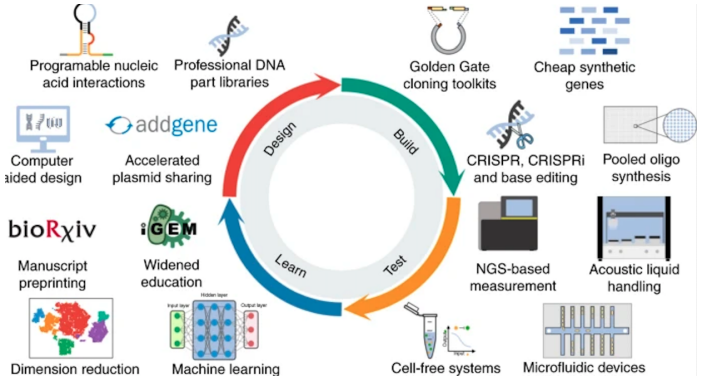




- traditional drug discovery process - time-consuming and costly often taking decades and billions of dollars
- AI streamlines this process by predicting the efficacy and safety of potential compounds with more speed and accuracy
- AI models analyze chemical databases to identify new drug candidates or repurpose existing drugs for new therapeutic uses
- companies - Insilco Medicine, Atomwise.

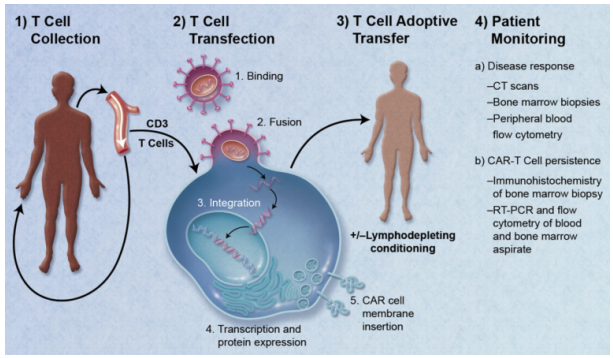
# Synthetic biology

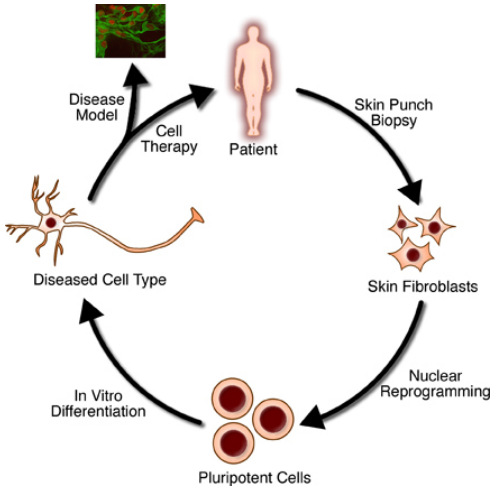
- use AI for gene editing, biomaterial production and synthetic pathways
- combine principles of biology and engineering to design and construct new biological entities
- AI optimizes synthetic biology processes from designing genetic circuits to scaling up production
- company - Ginkgo Bioworks uses AI to design custom microorganisms for applications ranging from pharmaceuticals to industrial chemicals



# Regenerative medicine

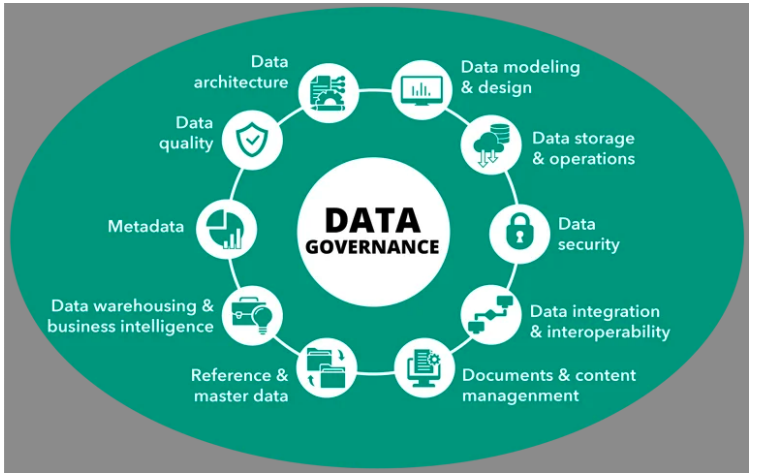



- AI advances development of stem cell therapies & tissue engineering
- AI algorithms assist in identifying optimal cell types, predicting cell behavior & personalized treatments
- particularly for conditions such as neurodegenerative diseases, heart failure and orthopedic injuries
- company - Organovo leverages AI to potentially improve the efficacy and scalability of regenerative therapies, developing next-generation treatments

# Bio data integration

- integration of disparate data sources, including genomic, proteomic & clinical data - one of biggest challenges in biotech & healthcare
- AI delivers meaningful insights *only when* seamless data integration and interoperability realized
- developing platforms facilitating comprehensive, longitudinal patient data analysis - vital enablers of AI in biotech
- company - Flatiron Health working on integrating diverse datasets to provide holistic view of patient health



# Silicon Valley's Cultural Engine of Innovation and Disruption

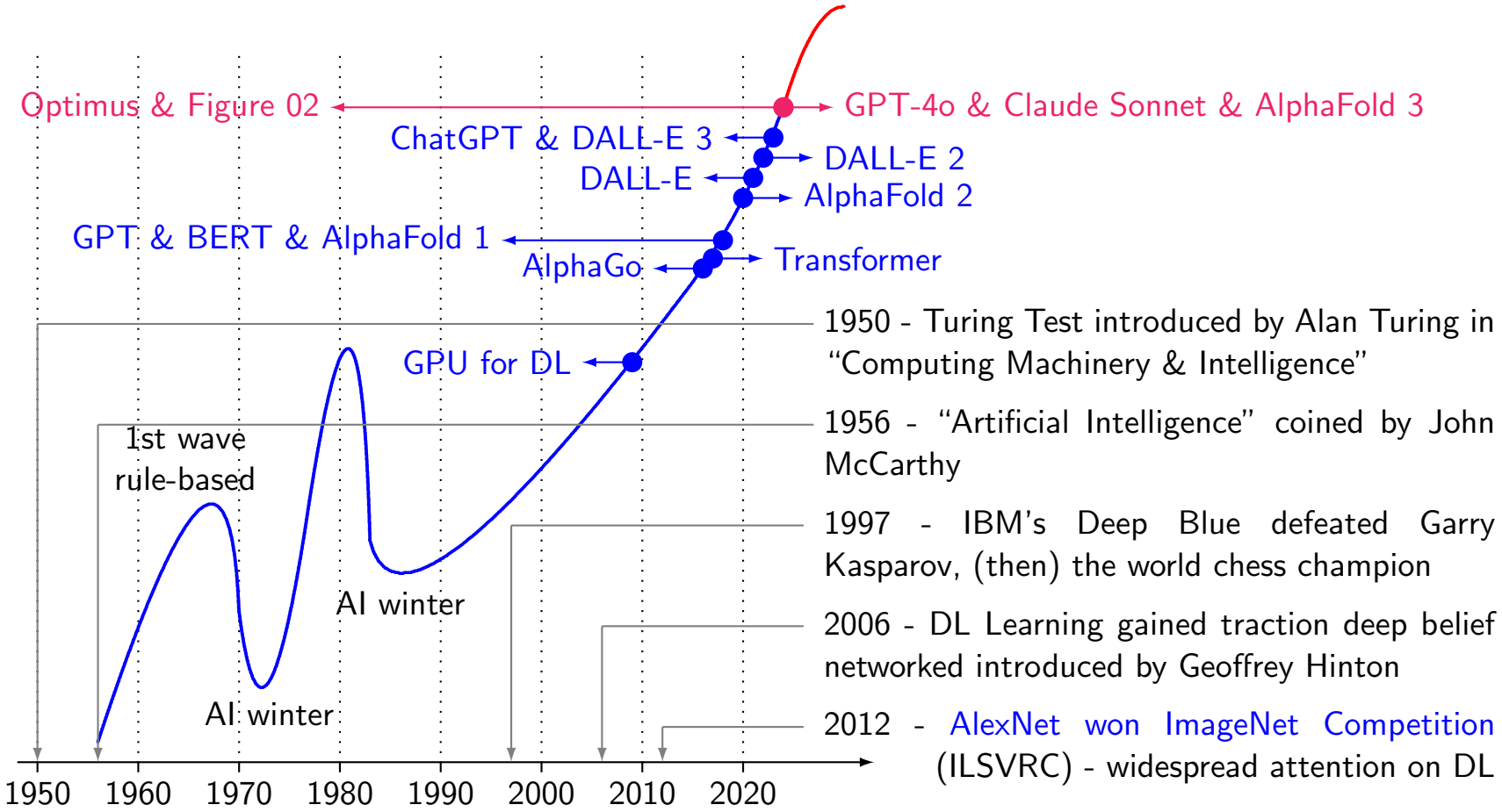
# My journey from Samsung & Amazon to Gauss Labs & Erudio Bio

- Samsung Semiconductor, Inc.
  - inception into industry from academia, the world's best memory chip maker!
- Amazon.com, Inc.
  - experience so-called Silicon Valley big tech culture and technology
  - set tone for my future career trajectory!
- Gauss Labs, Inc.
  - found & operate AI startup, shaping corporate culture & spearheading R&D as CTO
  - inherent challenges of Korean conglomerate spin-off startup - cultural constraints, over-capitalization, and leadership limitations
- Erudio Bio, Inc.
  - concrete & tangible bio-technology in addition to AI
  - great decisions regarding business development; business models, market fit, go-to-market (GTM) strategies based on lessons learned *in a hard way* ☺



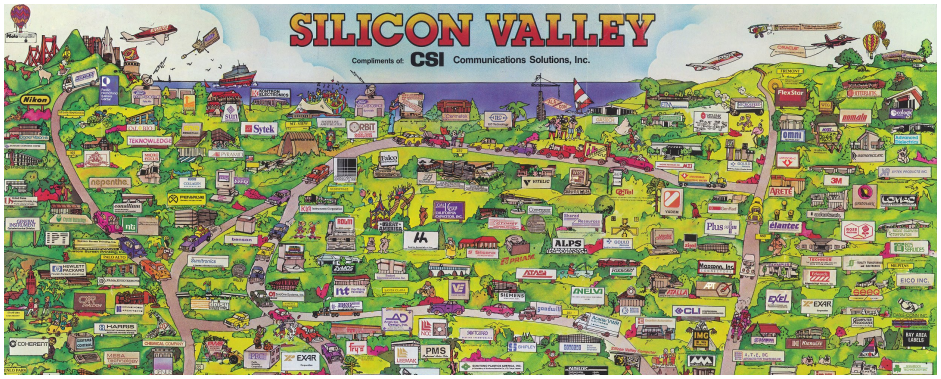


# Joining Amazon.com, Inc. at the inflection point of AI



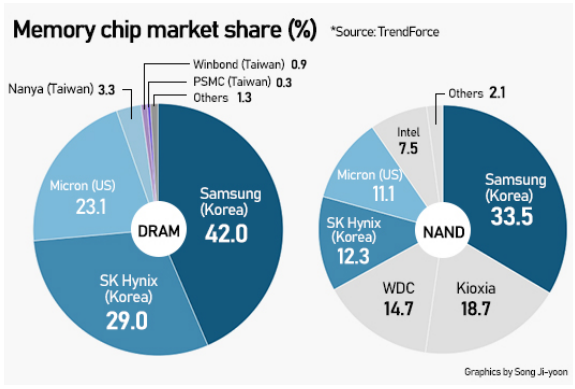
# Innovation ecosystem of Silicon Valley

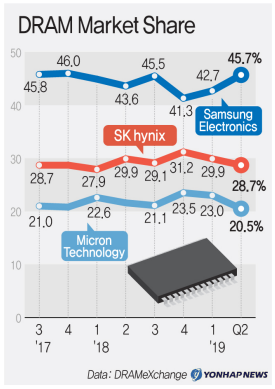
- key characteristics
  - risk-taking culture, *trust* in technology → *genuine* respect for engineers and scientists
  - easy access to huge capital - VCs, angel investors alike
  - talent density - engineers, researchers, scientists, entrepreneurs, PMs, TPMs, . . .
  - diversity, "collision density" of ideas
  - ecosystem of collaboration and competition - startups, academia, industry leaders
- what they mean for global big tech
  - set trends in AI, software & hardware (and or hence) product & industry innovation
  - act as testing ground for disruptive ideas




# Case study: Amazon - amazing differentiators of big techs

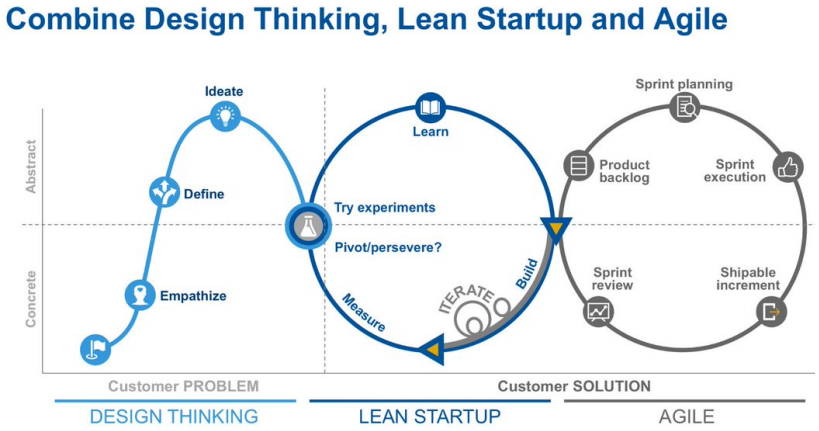
- Amazon's culture & leadership principles
  - customer obsession as driver of innovation
  - high standards & ownership culture, disagree & commit
  - bias for action and long-term thinking - sounds contradictory?
  - mechanisms like "two-pizza teams" & "Day One" for (or rather despite) scalability
- lessons for Korean corporations
  - applying customer-centric innovation in hardware & AI, *e.g.*, on-device AI
  - balancing agility with long-term R&D
  - *build / adapt / apply on the core strength of Samsung that no other company has!*

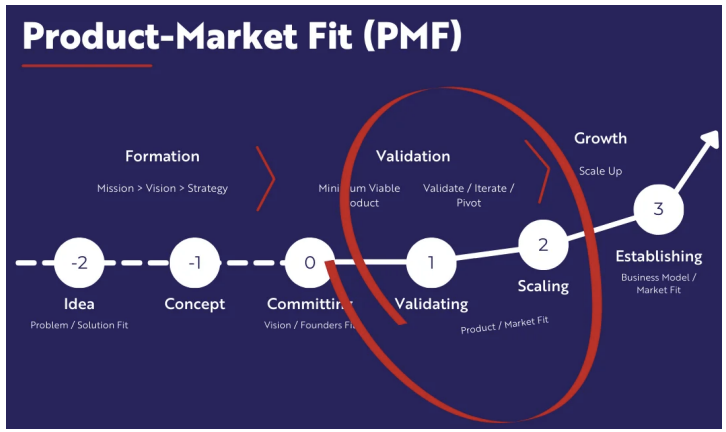



# Founding and scaling startups

- challenges
  - competence of and chemistry among co-founders crucial
  - technology & great team are *necessary*, but *not sufficient (at all!)* for success
  - business models, market fit, timing, agility, flexibility for pivoting / perseverance
- insight
  - importance of domain expertise in addition to AI
  - balancing innovation with good business decisions



# Bridging Silicon Valley & Korea

- cultural differences
  - risk appetite & failure tolerance
  - decision-making speed vs hierarchy
  - innovation vs execution focus
- opportunities for collaboration
  - leveraging Korea's manufacturing expertise with Silicon Valley's software/AI strengths
  - building global teams with diverse perspectives

# To be successful . . .

- embrace customer/market-centric mindset in innovation and for business decisions
- balance agility with long-term vision
- foster cross-cultural collaboration for global impact
- ((very) strategically and carefully) leverage AI to solve real-world industrial challenges


# Appendices

# K-PAI - Silicon Valley
# Privacy-Preserving AI Forum

# Silicon Valley Privacy-Preserving AI Forum (K-PAI)

- pioneering community of professionals dedicated to building privacy-preserving AI solutions, products, and systems
- comprehensive expertise across AI domains
  - biotechnology, healthcare, and medical research
  - industrial applications and data centers
  - cloud infrastructure, storage solutions, mobile technologies
  - customer service platforms, multi-agent systems
  - RAG implementations, vector databases, agentic AI frameworks
- vision
  - *shaping future where AI innovation and privacy protection go hand in hand*
- active community with homepage & KakaoTalk collaboration platform for members





# Our journey - forum history

- Nov-Dec 2024 - "The AI Strikes Back" & "Free Your Data"
  - Prof. Jung Hee Cheon (homomorphic encryption revolution)
- Jan 2025 - "The AI Knight Rises"
  - Sunghee Yun @ Erudio Bio on deep learning to flourishing societies
- Feb 2025 - "Silicon Citadel"
  - Chanik Park @ MangoBoost on AI data infrastructure
- Mar 2025 - "Blockchain Awakens"
  - Daejun Park @ a16z crypto on decentralized AI
- Apr 2025 - "Advancing Humanity"
  - Stanford Medicine team on bio/medical AI
  - co-hosting with K-BioX
- May 2025 - "The Autonomous Alliance"
  - Microsoft, GitHub, Uclone, SK Hynix on AI agents

# Our journey - forum history

- Jun 2025 - "Silicon Companions"
  - Altos Ventures on robotics & smart devices
- Aug 2025 - "The Human-Centric AI Revolution"
  - address legal and ethical issues related to AI
- Nov 2025 - "The AI Silicon Race"
  - Korea-US Innovation Leadership at K·ASIC


# Strategic partnerships & ecosystem

- *Perpetual Partnership with KOTRA Silicon Valley as Strategic Alliance*
- 2026 co-hosting partners
  - K·ASIC (Korea AI & IC Innovation Center)
  - K-BioX (biotech innovation)
  - KOTRA Silicon Valley (trade & investment)
  - Korean Consulate General, San Francisco (diplomatic support)
  - KABANC (Korean American Bar Association of Northern California - legal expertise)
- building bridges between Silicon Valley innovation and Korean institutional networks
- creating comprehensive support ecosystem: technical, legal, business, diplomatic

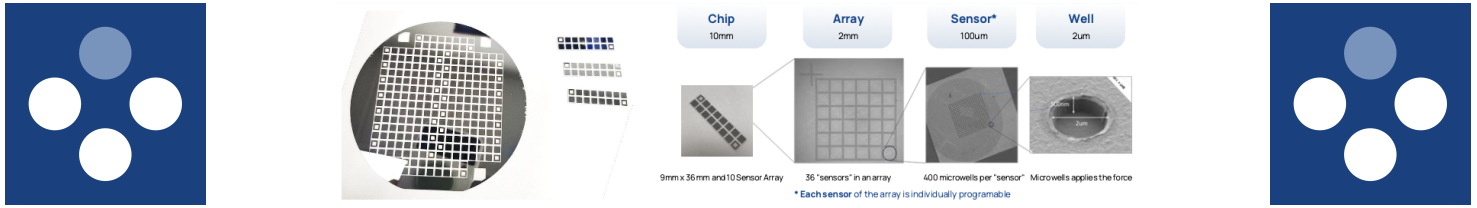
# Community & engagement

- membership requirements
  - attend 2+ K-PAI Forums to qualify
- member benefits
  - networking with AI professionals across all domains
  - knowledge sharing and collaboration opportunities
  - direct access to world-class speakers and experts
- forum format - 5pm-8pm, typically Wednesdays at premier Silicon Valley venues
- venues - Stanford, KOTRA, SK Hynix, Altos Ventures, K·ASIC, and more
- active community engagement and professional development

# Erudio Bio

# Powering AI-driven medicine with ground-truth binding data

- problems we solve
  - 90% of drugs fail in clinical trials due to poor early-stage prediction
  - multiplexed diagnostics suffer from false positives and cross-reactivity
- *Erudio Bio's Innovation*
  - VSA platform uses patented "dynamic force spectroscopy" to generate 1000x more high-quality binding data from single sample ($\sim 10\mu$L)
  - measuring not just presence, but *strength* and *kinetics* of molecular interactions
- *dual business model*
  - diagnostics - multi-cancer biomarker detection with clinical institutions & hospitals (Seoul National University Hospital Bundang, Keimyung University Dongsan Hospital)
  - drug discovery - bioTCAD$^{\mathrm{TM}}$ platform providing ground-truth labels to train & validate pharma AI models, reducing preclinical cycles



# Validated technology, proven team, clear path to market

- validated impact
  - *$1M Gates Foundation Grant* (2025) to democratize drug development for global health
  - partnerships with top research institutions (KRIBB, KAIST)
- unique team - *Stanford-trained founders* combining
  - semiconductor TCAD expertise & force spectroscopy innovation (20+ years)
  - AI & optimization leadership (Samsung, Amazon, SK hynix, Gauss Labs)
- market entry
  - *Korea → Asia hub & US* strategy with 2026 regulatory milestones and expanding pharma partnerships

Gates Foundation

# Biological assays struggle with scale & accuracy

# Data is expensive

- so we make decisions with *incomplete* picture
- status quo
  - limited, small-scale testing confirms diagnosis
  - outcome only as good as doctor's ability to determine which tests, limiting the picture
  - cross reactivity prevents larger scale testing
- Erudio creates
  - *comprehensive, large-scale* testing will drive diagnosis without assumptions
  - increased scale enables enhanced scientific discovery leading to
    - *better patient care*
    - *reduced time to diagnosis*
    - *cost reduction*



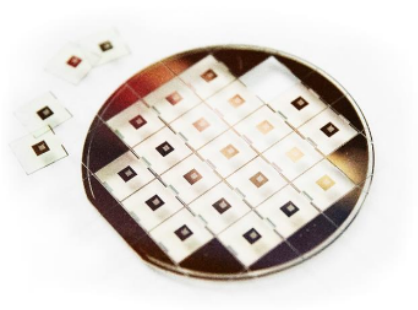

# Erudio Bio starting Revolution

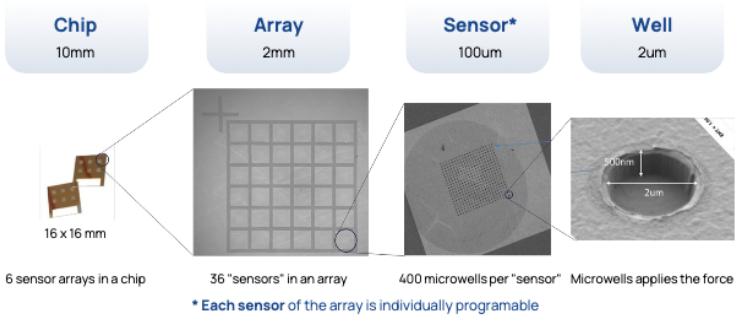
# Erudio Bio starts revolution with Gates Foundation's support

- more data
  - comprehensive data from single biological sample
  - multiplexed analysis of nucleic acid, protein, cells, and more!
  - *multi-omic platform*
- actionable data
  - combined quality score from all data sources for comprehensive & conclusive assessment
- earlier data
  - complete data early to drive accurate decision making

# Versatile Smart Assay (VSA) Platform

# VSA technology





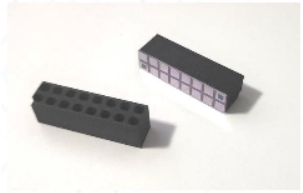

- generates *1000x more data* than the prevailing technology
  - scalable multi-omic microarray sensor
- *21 patents* in US, Canada, China, and Europe
- indicates how good the data is in real time
  - patented "dynamic force spectroscopy" and "powerful Bayesian inference" method provides our data *quality score* to know their accuracy for actionable data
- AI software extracts a detailed, interpretable picture for quick diagnosis
  - leads to *AI knowledge discovery* resulting in *data-driven diagnosis*

# Enabling comprehensive data acquisition

- antibodies - versatile tools in biology
  - can engineer to target virtually *anything* we want
  - problem
    - indiscriminate interactions severely limits use of antibodies in multiplex formats
    - error-prone results due to non-specific binding
- solution - comprehensive data with *dynamic force spectroscopy*
  - comprehensive binding strength to distinguish specific from non-specific binding
  - *quality score* discerns noise from useful data to enable multiplexing



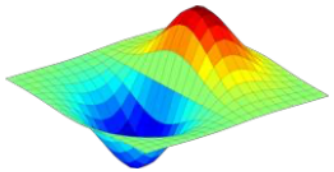

# Erudio Bio's business models



Consumable chip & flowcell


Instrument

Consumable reagent kit



Software AI/ML & SaaS

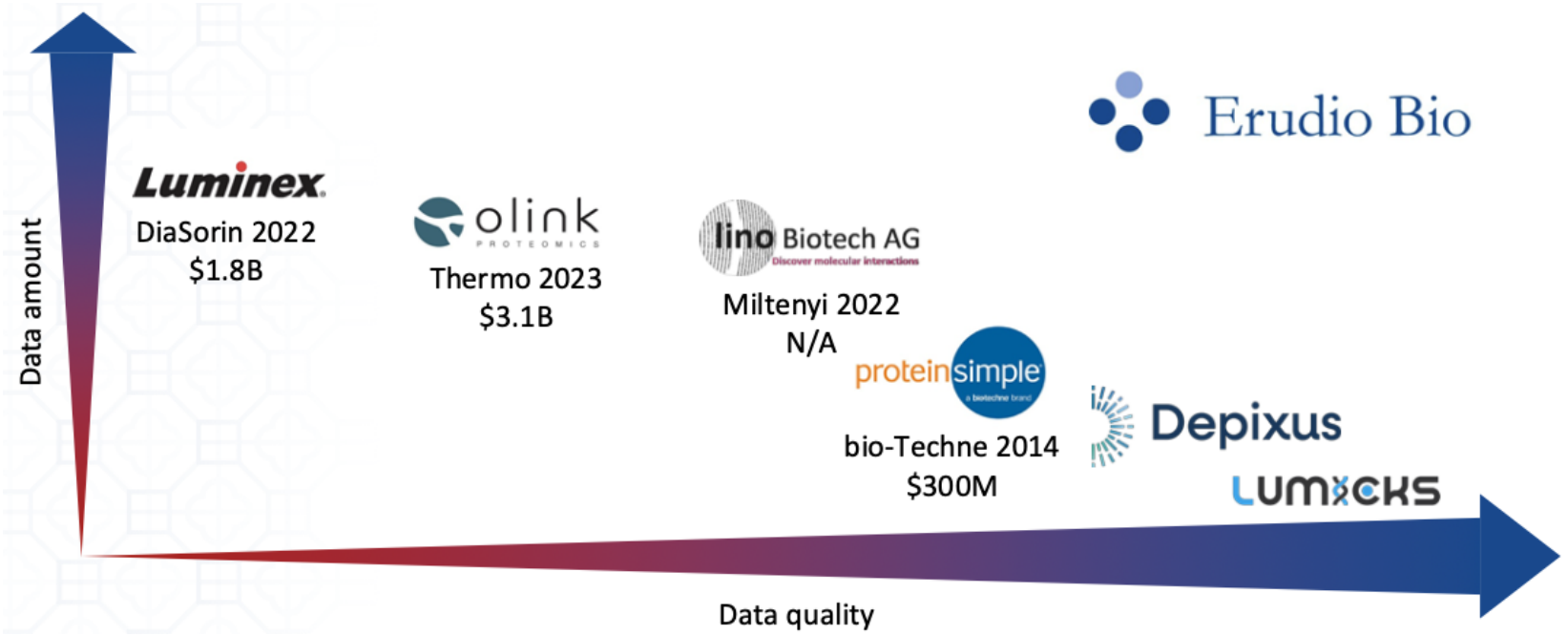
- VSA platform
  - instrument - recurring revenue with high margin
  - modular licensable software - AI based data interpretation and feature extraction
- SaaS
  - subscription based pre-validation of reagent database
  - AI feature extraction and knowledge discovery

# Milestones

# Erudio Bio key milestones

- IP developed at Stanford School of Medicine under mentorship by father of microarrays - Dr. Ronald Davis
- data vetted by Analog Devices ($115B market cap semiconductor company), Harvard Medical School & Massachusetts General Hospital
- commercial partnership with Analog Devices to manufacture at scale
- JDA with Shanghai General Hospital's National Clinical Research Center for Eye Diseases
- JDA with Seoul National University General Hospital (SNUBH) for multiplexed cancer biomarker detections
- partnerships with
  - Keimyung University Dongsan Hospital, KRIBB, KAIST NanoFab, Lulumedic

# Competitive landscape




- Erudio Bio
  - *only company* providing both high quality data and large data output
- efficient workflow integration facilitates customer adoption

# Erudio Bio engaging with Customers

# Validating out technology in clinical settings


- joint development agreement signed with *Shanghai General Hospital's National Clinical Research Center for Eye Diseases*
  - co-develop multiplexed diagnostic test for uveitis
  - addresses low sample volume and diagnostic inaccuracies
- globally unique partner, customer so large that it is a market by itself
  - 45 large hospitals with 83M patient visits per year
  - standard of care for smaller hospitals in Shanghai to access additional 280M patients

# Clinical validation to market leadership in Korean preventive care


- market opportunity - *preventive care* is foundation of Korean healthcare
  - $\sim$ 15M health screenings performed in S. Korea testing
  - ideal market segment for Erudio Bio's multi-omic multiplexed VSA platform
  - multi-billion dollar addressable market for multi-cancer early detection
- joint development agreed with *Seoul National University Hospital Bundang (SNUBH)*
  - premier hospital of national importance
  - enabling expansive proactive health assessment for efficient health care system
  - collaboration with target of *multiplexed cansor biomarker medical equipment business*

# Teams

# Team & advisory board

- team
  - Sunghee Yun, Ph.D. (CTO) - AI, optimization, business development, software
  - Kee-Hyun Paik, Ph.D. (CEO) - chip, microfluidics, instrumentation
  - Susanne Baumhueter, Ph.D. - biology, immunology, project management
  - Leon Chen, MBA, CFA (COO) - business development, product, operations
  - Jin Young Huh (CLO) - chief legal officer, business development in Korea
- advisory board
  - Michael Cola - CEO of AEVI Genomic Medicine ($62B sales to Takeda)
  - Tim Germann - CCO of Carterra Bio
  - Karyn Eliot - retired CIA Sr. Executive
  - Ronald W. Davis - Director of Stanford Genome Tech Center ($15B+ exits)
  - William J. Greenleaf - Prof. Genetics and Applied Physics, Stanford University

# Gates Foundation Grant

# Erudio Bio wins $1M Gates Foundation Grant - scaling bioTCAD

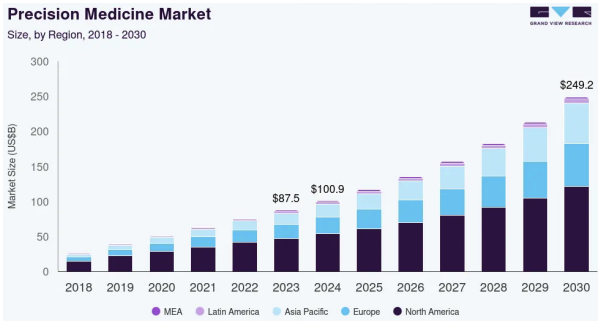
Gates Foundation

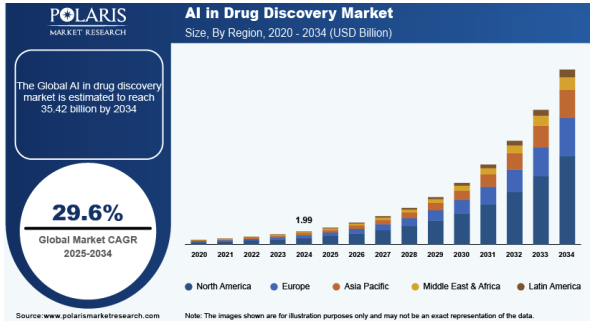

- $1M Grant Award (August 2025)
  - Gates Foundation recognizes Erudio Bio's potential to transform drug development for global health
- mission alignment - democratizing medicine by making preclinical drug design faster, yet reliable & accessible
  - lowering development costs for diseases affecting low- and middle-income countries (LMICs)
  - addressing the 90% clinical trial failure rate that drives up drug costs
- funded project - scale bioTCAD$^{\mathrm{TM}}$ platform to generate ground-truth binding datasets
  - expand force spectroscopy measurements across high-burden disease targets
  - train AI models with kinetics-resolved binding data (on/off rates, unbinding forces)
  - enable pharma/biotech to prioritize candidates earlier with higher confidence

# The Trillion Dollar Opportunities

# The market is real - and enormous

- AI in biotech/pharma market size
  - AI in drug discovery - $4B (2023) $\rightarrow$ $50B+ by 2034 (Global Market Insights)
  - AI diagnostics market - $1.2B (2023) $\rightarrow$ $5-12B by 2030
  - precision/personalized medicine - $80B (2023) $\rightarrow$ $230B by 2030
  - synthetic biology - $15B (2023) $\rightarrow$ $100B by 2032
  - *combined TAM approaching $1T by mid-2030s* - conservative estimate
- why biotech AI multiples exceed pure software AI
  - software AI competes on marginal cost $\rightarrow$ commoditizes fast
  - biotech AI - irreversible IP - novel molecules, validated biomarkers, proprietary assay
  - every successful clinical trial is a data moat that cannot be reverse-engineered

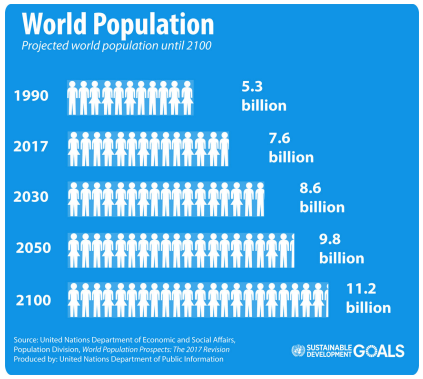


# Macro tailwinds amplifying these opportunities

- aging global population - over 2.1 billion people over 60 by 2050 (UN DESA)
  - cancer, neurodegeneration, cardiovascular disease scale accordingly
- post-COVID regulatory acceleration
  - FDA's AI/ML action plan, EMA adaptive pathways
- government prioritization - US CHIPS & Science Act includes biotech
  - Gates Foundation, NIH, BARDA deploying billions in non-dilutive capital
- historical parallel
  - genomics wave (Human Genome Project) unlocked $1T in economic value
  - *AI-biotech convergence is at least one order of magnitude larger*

Gates Foundation



# Three exponential curves converging now

- *curve 1 - LLMs & genAI*
  - language models operating natively on biological "languages" - protein sequences (ESM-2, ESMFold), SMILES molecular notation, genomic sequences
  - generative AI can propose novel drug candidates - Insilico Medicine's INS018_055 went from AI-generated candidate to Phase II in 4 years vs industry avg of 10–15 yrs
  - multi-modal AI integrating imaging, omics, and clinical notes simultaneously
- *curve 2 - biochemical & biological breakthroughs*
  - AlphaFold 3 (2024) - extends beyond proteins to DNA, RNA, small molecules, and their interactions — the full drug-target interface
- *curve 3 - data availability*
  - DNA sequencing cost reduction, INSDC database, electronic health records digitized, wearables & continuous monitoring
- *The rarity of this moment - all three curves peaking simultaneously is historically unprecedented*

# Cross-domain inevitabilities - The Technical Moat

- core insight
  - fundamental mathematical structures recur across seemingly unrelated domains
  - recognizing these - strategic advantage - optimization $\longrightarrow$ biological energy landscapes
  - protein folding is fundamentally an energy minimization problem over high-dimensional conformational space
- information theory $\longrightarrow$ cellular signaling
  - mutual information and channel capacity concepts (Shannon, 1948) map directly onto how cells encode and transmit signals through biochemical cascades
  - LLM training optimizes cross-entropy loss over token distributions
  - cellular gene regulatory networks optimize analogous information-theoretic objectives over transcription factor binding distributions
  - researchers who understand why transformers work can transfer those architectural intuitions to biological sequence modeling

# What separates Unicorn Potential from incremental progress

- *dimension 1 - platform vs point solution*
  - point solution - AI model predicting one biomarker for one cancer type - narrow addressable market & low defensibility
  - *platform* - technology applicable across multiple disease areas, multiple biomarker classes, multiple assay modalities — TAM compounds with each new application
  - litmus test - "Can this technology be redirected to new disease areas in a few months without rebuilding from scratch?"
- *dimension 2 - is AI load-bearing or decorative?*
  - "AI-washing" - ML used for marketing positioning, not scientific differentiation
  - *load-bearing AI* - the AI component creates a result impossible or uneconomic to achieve otherwise
    - $e.g.$, Erudio Bio's dynamic force spectroscopy + AI detecting cancer biomarkers at concentrations below conventional immunoassay thresholds
- *dimension 3 - pathway to clinical and regulatory reality*
  - computational elegance that cannot survive contact with clinical data is worthless

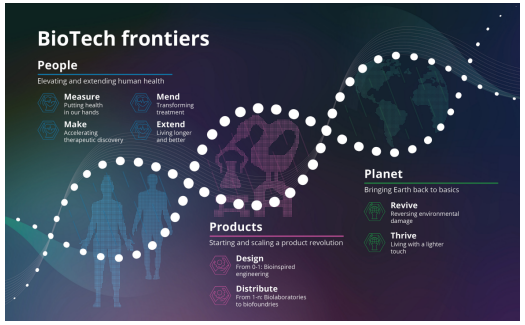
  - regulatory strategy must be designed into product from day one - not retrofitted after technical development
  - FDA's 510(k) vs De Novo vs PMA pathways have entirely different clinical evidence requirements — choice of pathway is a strategic decision made at founding
  - *hospital partnerships ($e.g.$, SNUBH) not just validation - they are the pipeline for clinically grounded training data!*
- *dimension 4 - proprietary data moat*
  - the most durable competitive advantage in AI-biotech is *the data no one else can access or replicate*
  - $e.g.$, proprietary assay platforms generating novel measurement types, exclusive hospital partnerships, patient cohorts with longitudinal follow-up, rare disease registries
- *dimension 5 - human welfare at the center; not a constraint, but a strategic asset!*
  - mission alignment with patient outcomes unlocks - NIH/Gates/BARDA non-dilutive funding, academic medical center partnerships, favorable regulatory posture, and - increasingly - LP mandates in impact-oriented VC funds

# Investment landscape & white spaces

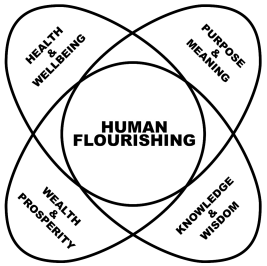
- where capital is concentrating - crowded but justified
  - *AI-native drug discovery platforms* - Insilico Medicine, Recursion Pharmaceuticals, Exscientia, Schrödinger - well-capitalized, public or late-stage
  - *protein engineering and design* - Absci, Generate Biomedicines, Cradle - foundation model approach to antibody and enzyme design
  - genomics interpretation - large language models trained on genomic sequences (Evo, Nucleotide Transformer)
  - risk - these spaces are getting crowded - differentiation increasingly difficult, capital efficiency under pressure
- underinvested white spaces - higher risk-adjusted opportunity
  - *AI-native diagnostic assays* - most diagnostic AI is retrofitted onto existing assay platforms - companies building AI-first measurement modalities have structural advantages in sensitivity, cost, and data proprietary
  - AI for *rare and neglected diseases* - Gates Foundation, Wellcome Trust, BARDA actively funding — orphan drug designation provides 7-year market exclusivity, priority review vouchers worth $\$100\text{M}+$ on the market

# The Rare Entrepreneur This Moment Demands

- three simultaneous operating modes – all required
  - *deep technologist* - must understand inner workings well enough to know what AI can and cannot do & *spot when a competitor's technical claim is hollow*
  - *business strategist* - must navigate regulatory complexity, reimbursement strategy, partnership structuring, and capital allocation under uncertainty simultaneously
  - *advocate for human flourishing* - must hold *patient outcome as the north star* – not as marketing message, but as decision criterion resolving conflicts between speed, cost, and scientific rigor

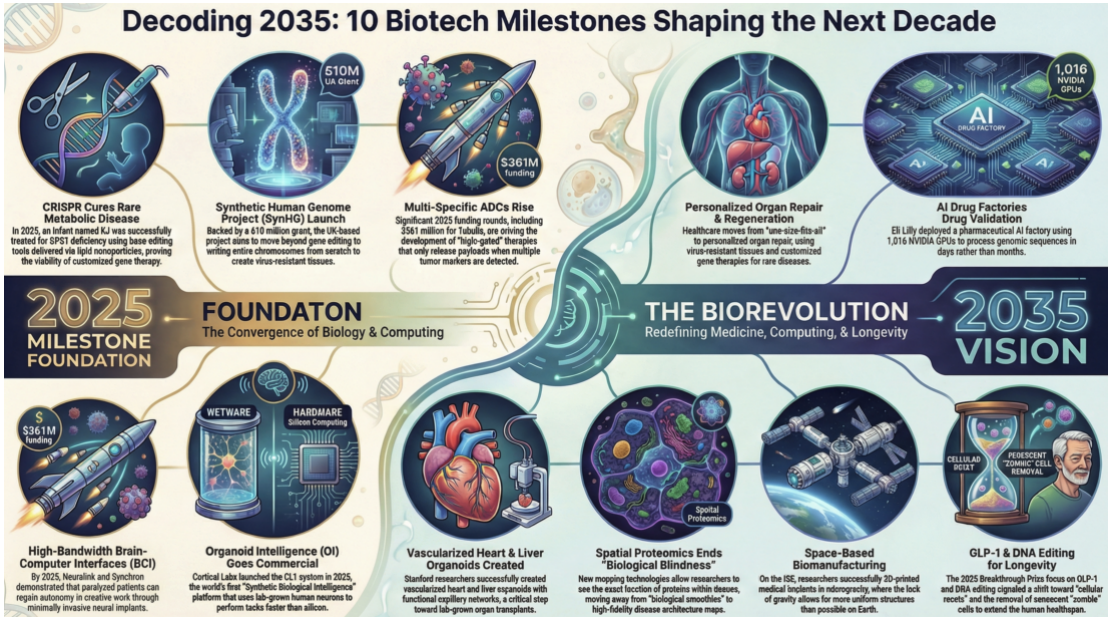






# What MIT GSW audience in this room should take away

- *the convergence is real, the market is massive, and the technical foundations are now mature enough to build on*
- the scarcest resource is not capital, nor compute, nor biological data – it is *founders holding the full stack - theory, biology, engineering, clinical reality, and human purpose*
- companies that will define human health in 2035 being founded in the next 24 months



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- CEOs, CTOs, CFOs, COOs, CMOs & CCOs @ startup companies in Silicon Valley
- VCs on Sand Hill Road - Palo Alto, Menlo Park, Woodside in California, USA

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# Thank You