

Tehisintellekt: sissejuhatus



- Orientatsioon
- Tehisintellekti tüübid ja valdkonnad
- Hübriidsüsteemid
- Robotid
- Filosoofia ja prognoosid
- Keelemudelite rakendamise hetkeseis

Orientatsioon

Marc Andreessen:

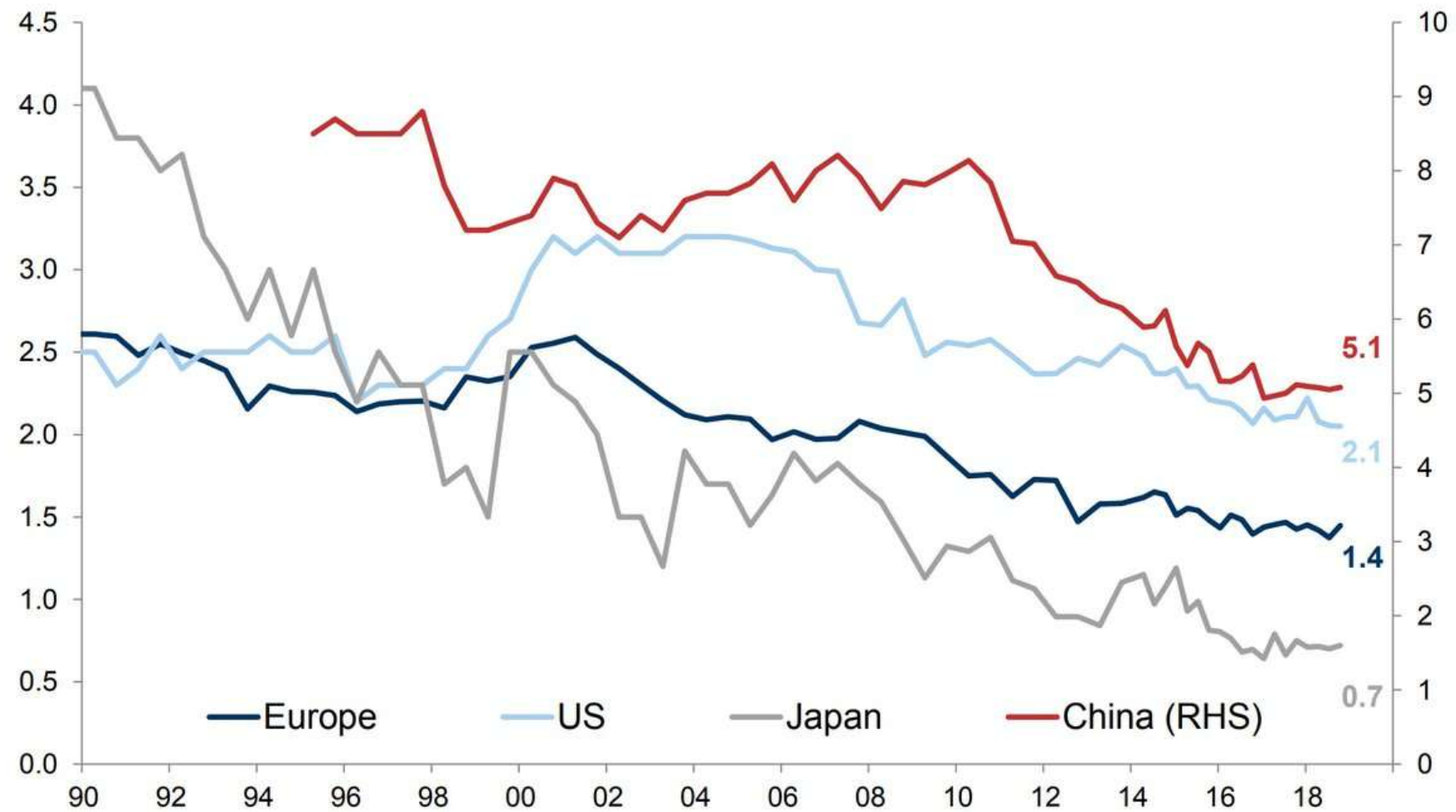
Software Is Eating the World

Paradoks:

IT ei ole sel sajandil globaalmajandust oluliselt efektiivistanud: vastupidi, GDP kasvutempo on pikaajalises langustrendis.

Exhibit 8: Long-term real global GDP growth forecast is at a historical low

Long-term (6-10y) GDP growth from Consensus Economics



Source: Consensus Economics, Goldman Sachs Global Investment Research

IT globaalne probleem ja lootus

Tööprotsesside automatiseerimine on siiani olnud väga raske: tarkvara ehitamine on olnud aeglane ja kallis.

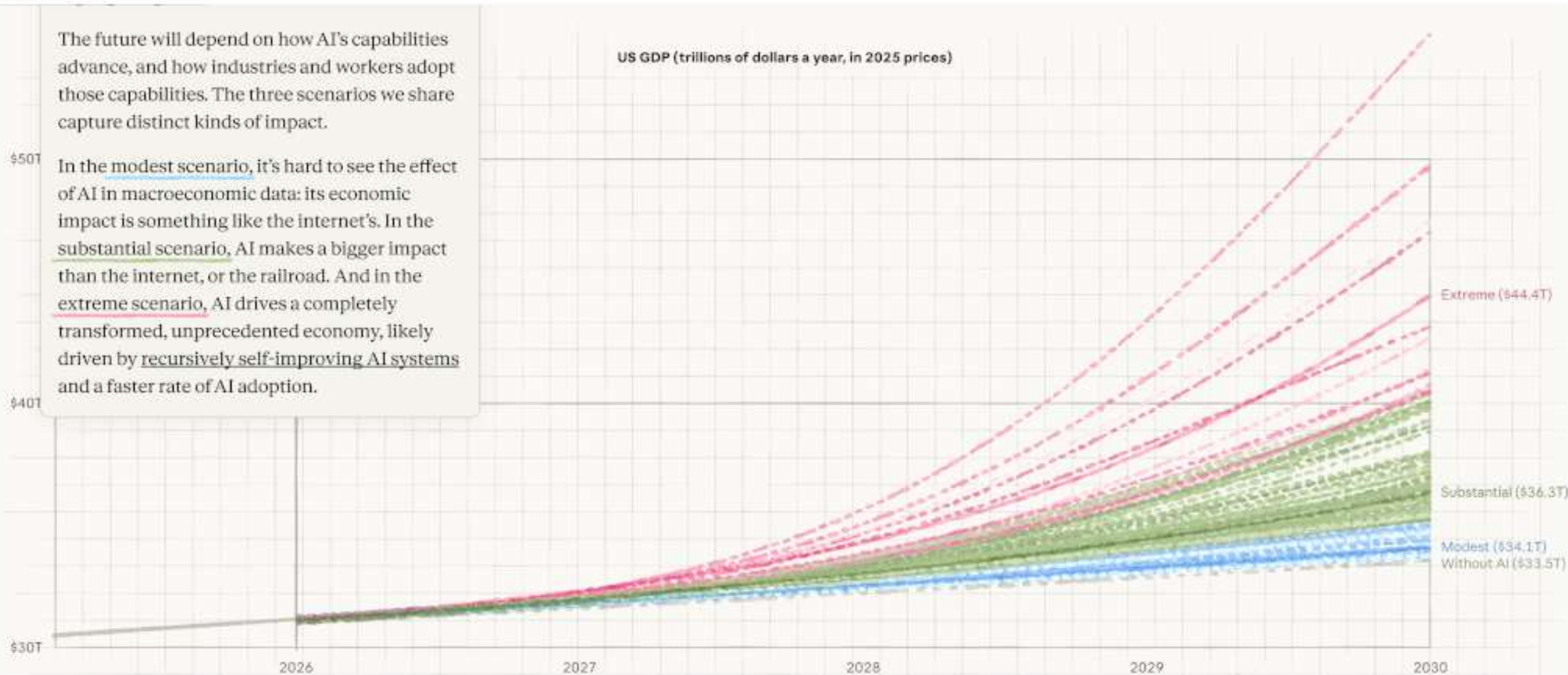
Praegu aitavad tehismõistuse tehnoloogiad mh palju efektiivsemalt ehitada automatiseerimis-tarkvara.

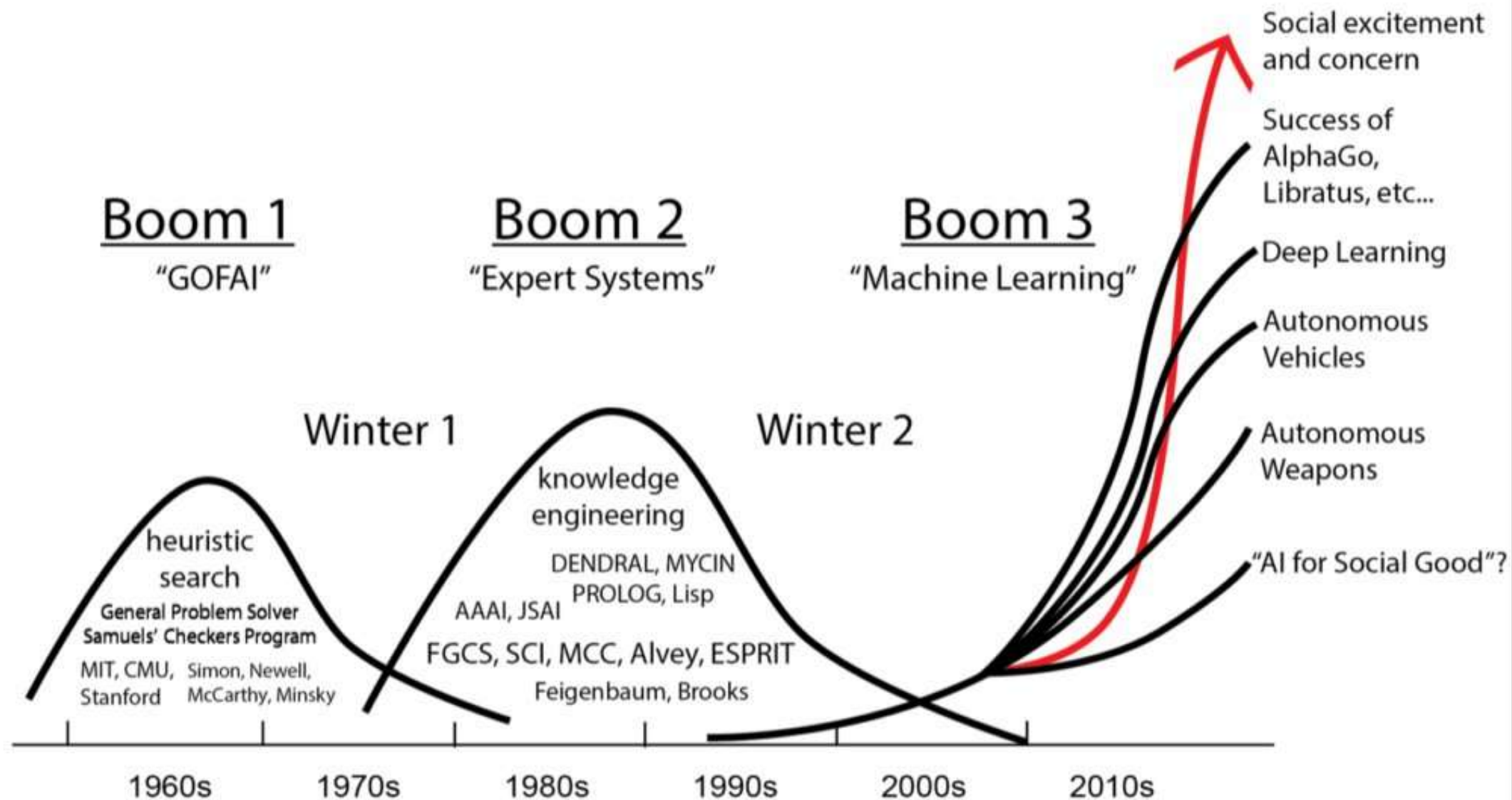
Oleme protsessis, kus tava-tarkvara ehitamise asemel saab üha enam töid automatiseerida otse LLMdele antavate ülesannete + taustainfo abil.

Anthropic: optimistlikud ennustused 2026 september

The future will depend on how AI's capabilities advance, and how industries and workers adopt those capabilities. The three scenarios we share capture distinct kinds of impact.

In the modest scenario, it's hard to see the effect of AI in macroeconomic data: its economic impact is something like the internet's. In the substantial scenario, AI makes a bigger impact than the internet, or the railroad. And in the extreme scenario, AI drives a completely transformed, unprecedented economy, likely driven by recursively self-improving AI systems and a faster rate of AI adoption.





Tehisintellekti tüübid ja valdkonnad

Kolm suuremat A.I. meetodite valdkonda:

- **Simulatsioonid**

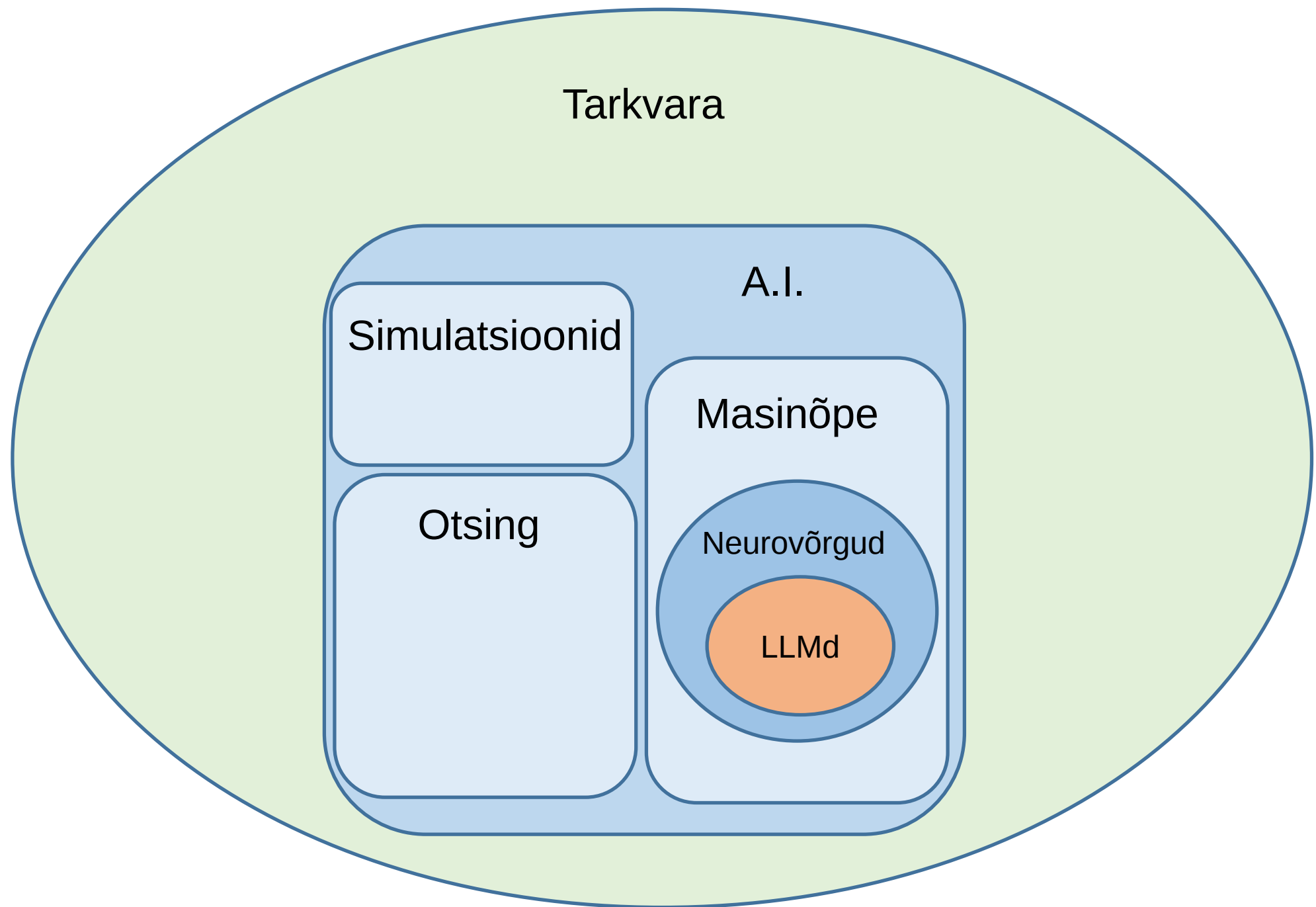
tegevuse tulemuste ja valdkondliku reaalsuse kiire prognoosimine a la FPS, digitaalsed kaksikud

- **Otsing** (sümbol-A.I.)

paljude variantide seast a la malemäng, matemaatikaõestus, teekonna optimeerimine, planeerimine jne.

- **Masinõpe** (“neuro”-A.I.)

näidetelt üldistamine, a la objektide äratundmine pildilt, deepfake, aegridadelt ennustamine, suured keelemudelid (LLM) GPT, LLaMA, Claude jne.



CAPTURE THE FLAG!

PRESS  TO CHANGE MISSION

22:37



GEORGE JOINED THE GAME

LEONARDO NICE GUY

73%

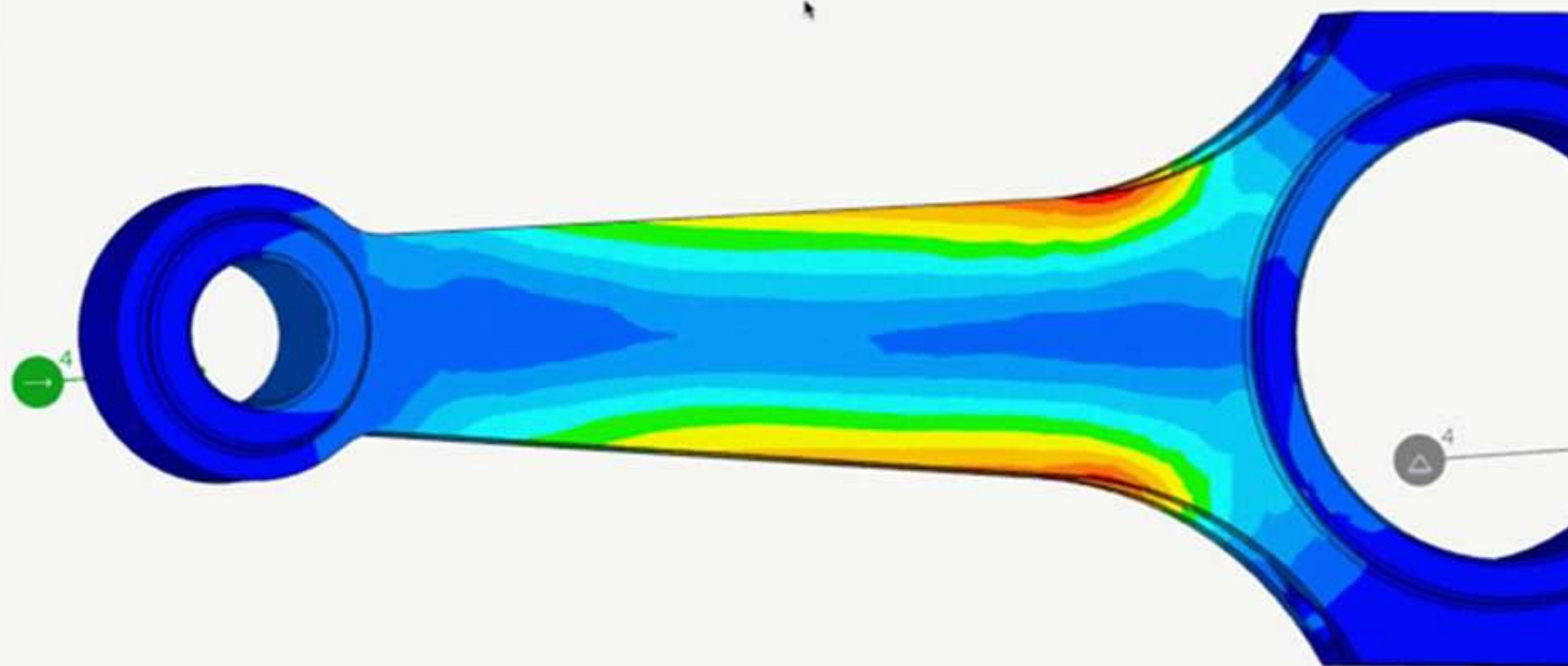
0%

73%

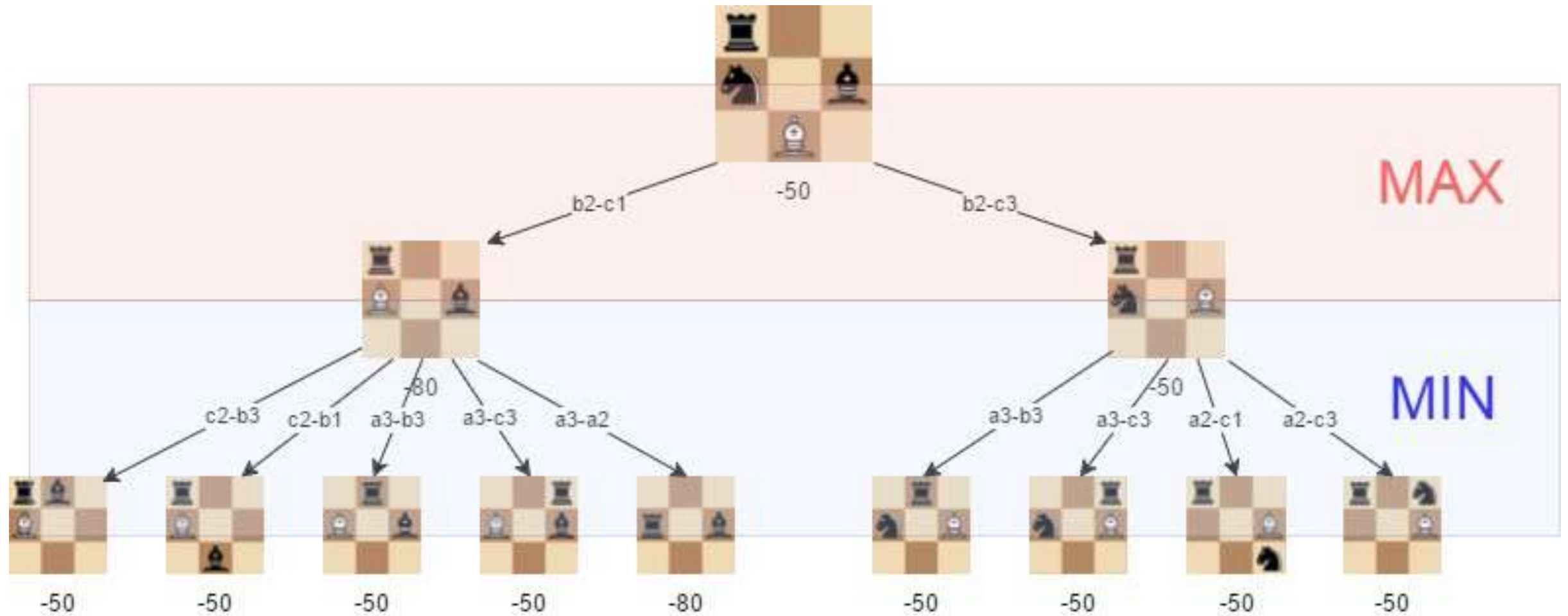
12 23

- GEOMETRIES (1)
- SIMULATIONS
 - Static
 - Geometry
 - Contacts (0)
 - Connectors (0)
 - Element technology
 - Model
 - Materials
 - Steel
 - Boundary conditions
 - Fixed support 1
 - Force 2
 - Numerics
 - Simulation control
 - Result control
 - Mesh
 - Simulation Runs

Job status



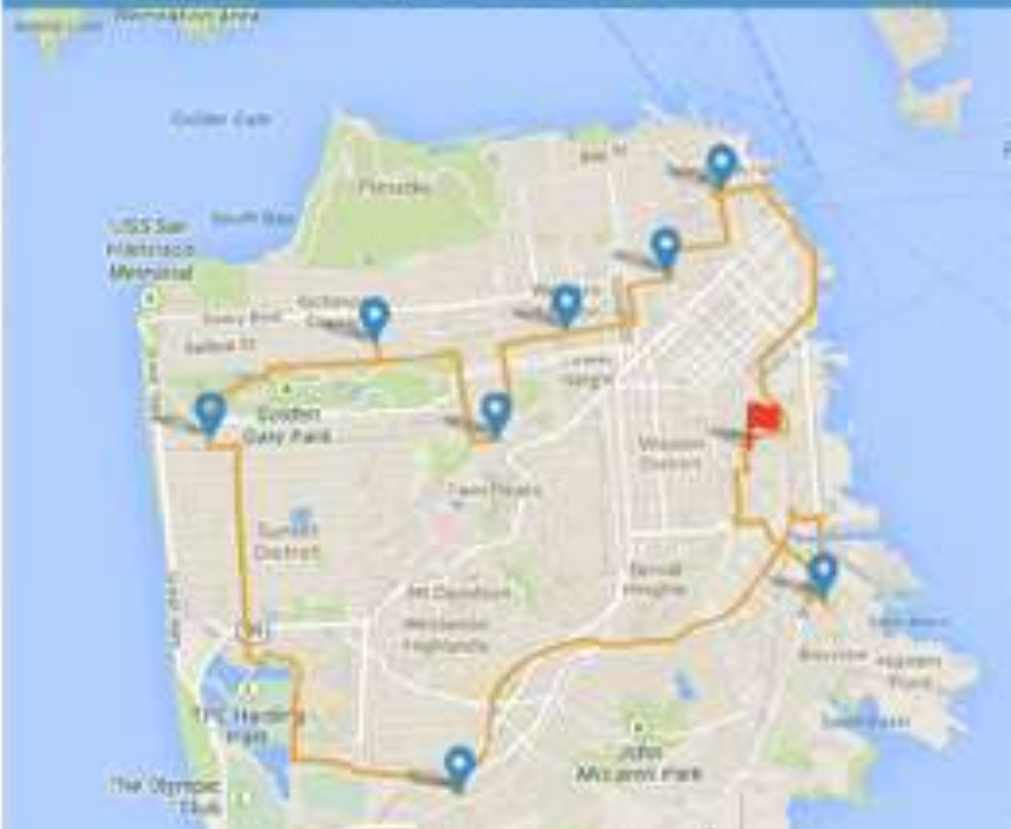
Otsing variantide hargnevast puust

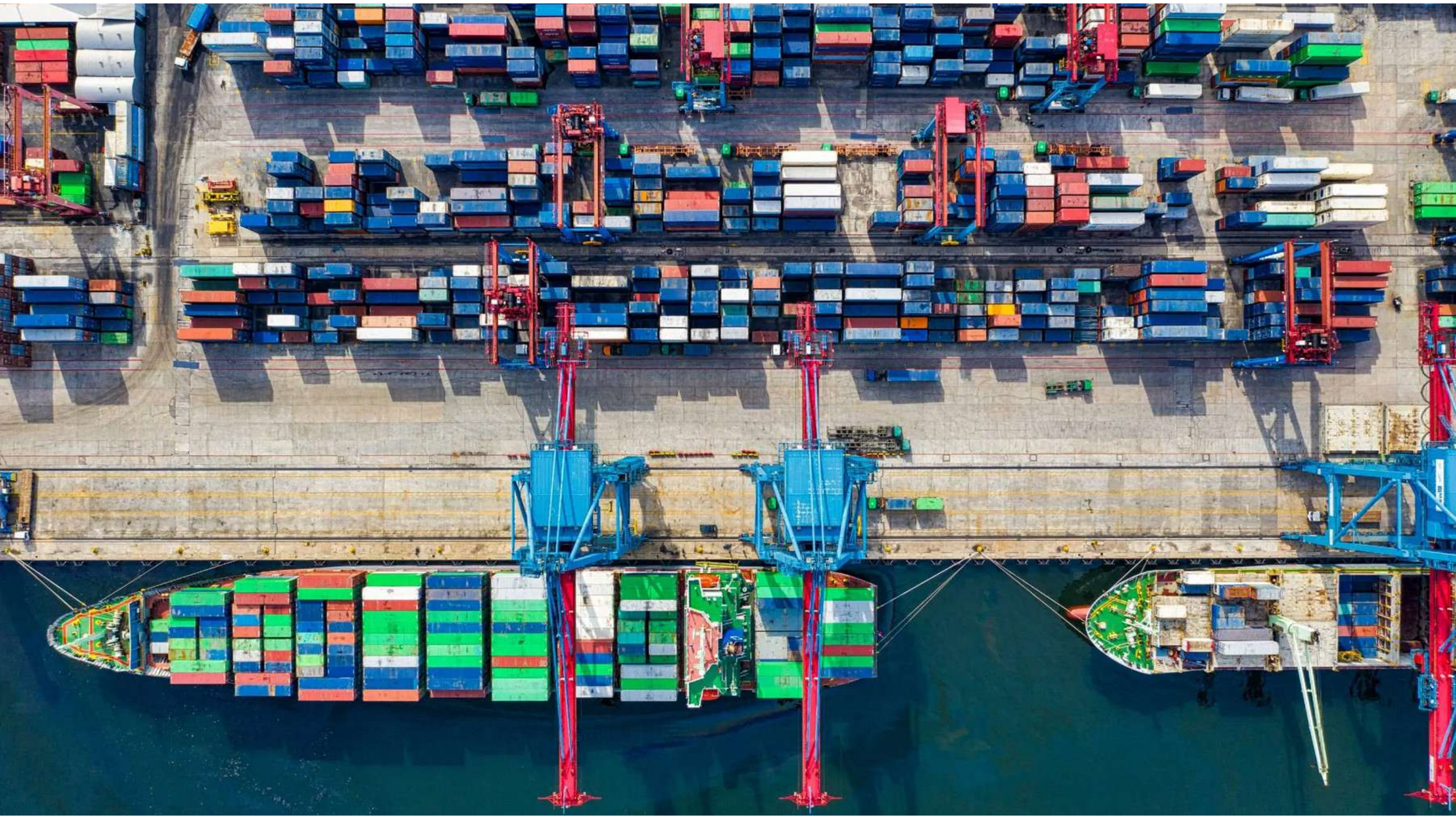


Unoptimized Route

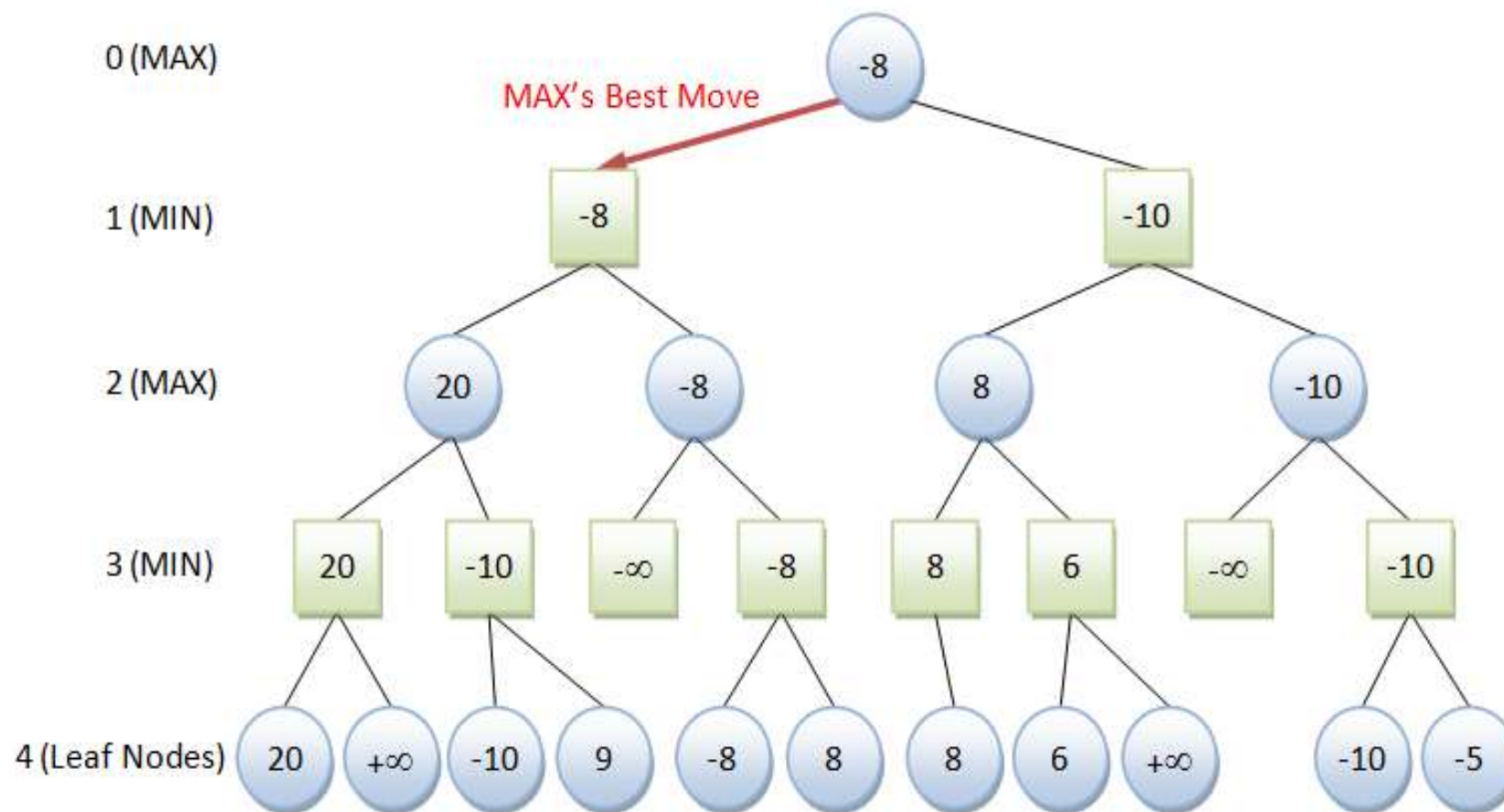


Optimized Route

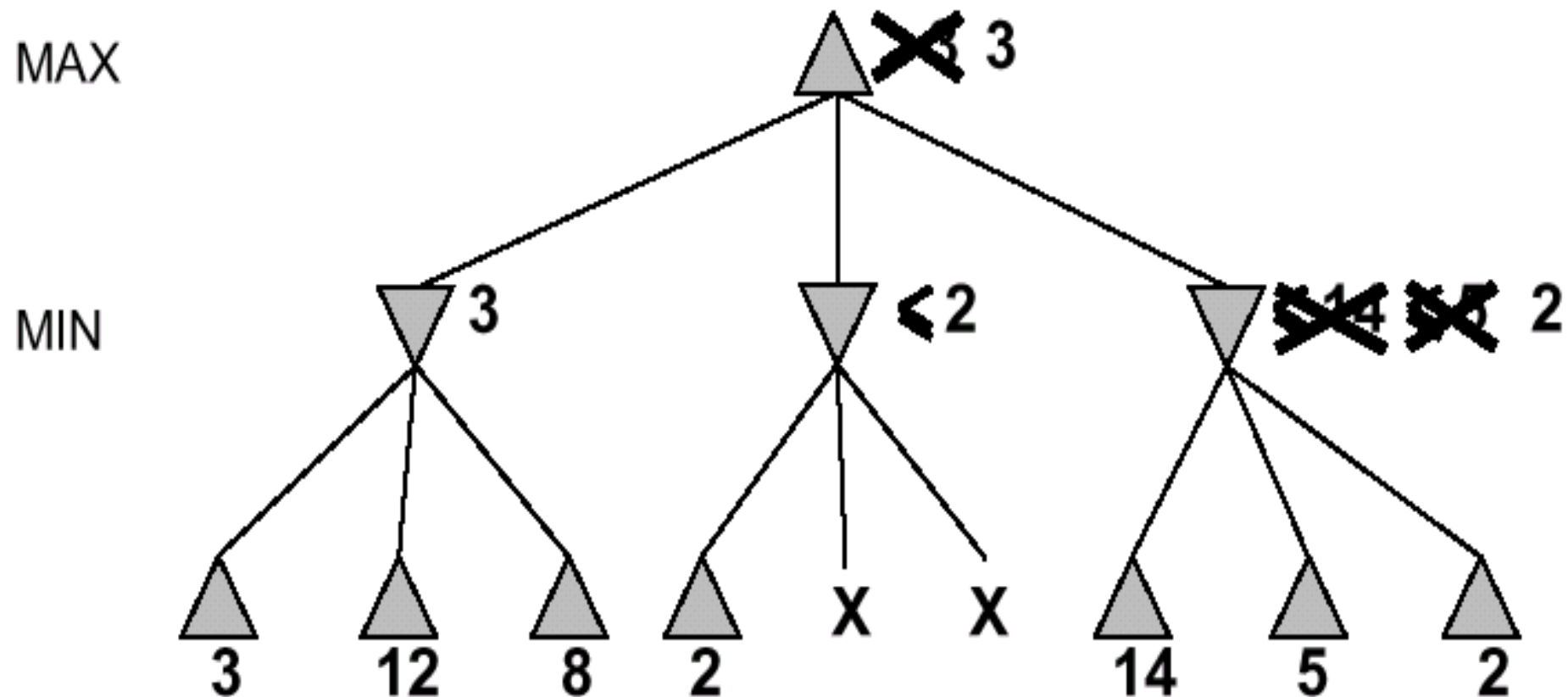




Otsing mängupuust: **minimax** algoritm



Üks esimesi universaalseid otsinguoptimeerimise meetodeid: alpha-beta



Rule-based systems, including expert systems

Given

`father(john,pete)` John is a father of Pete

`father(pete,mark)` Pete is a father of Mark

`father(X,Y) & father(Y,Z) => grandfather(X,Z)`

`father(X,Y) & mother(Y,Z) => grandfather(X,Z)`

If X is a father of Y and Y is a father/mother of Z, then X is a grandfather of Z

`grandfather(X,Y) => grandson(Y,X)`

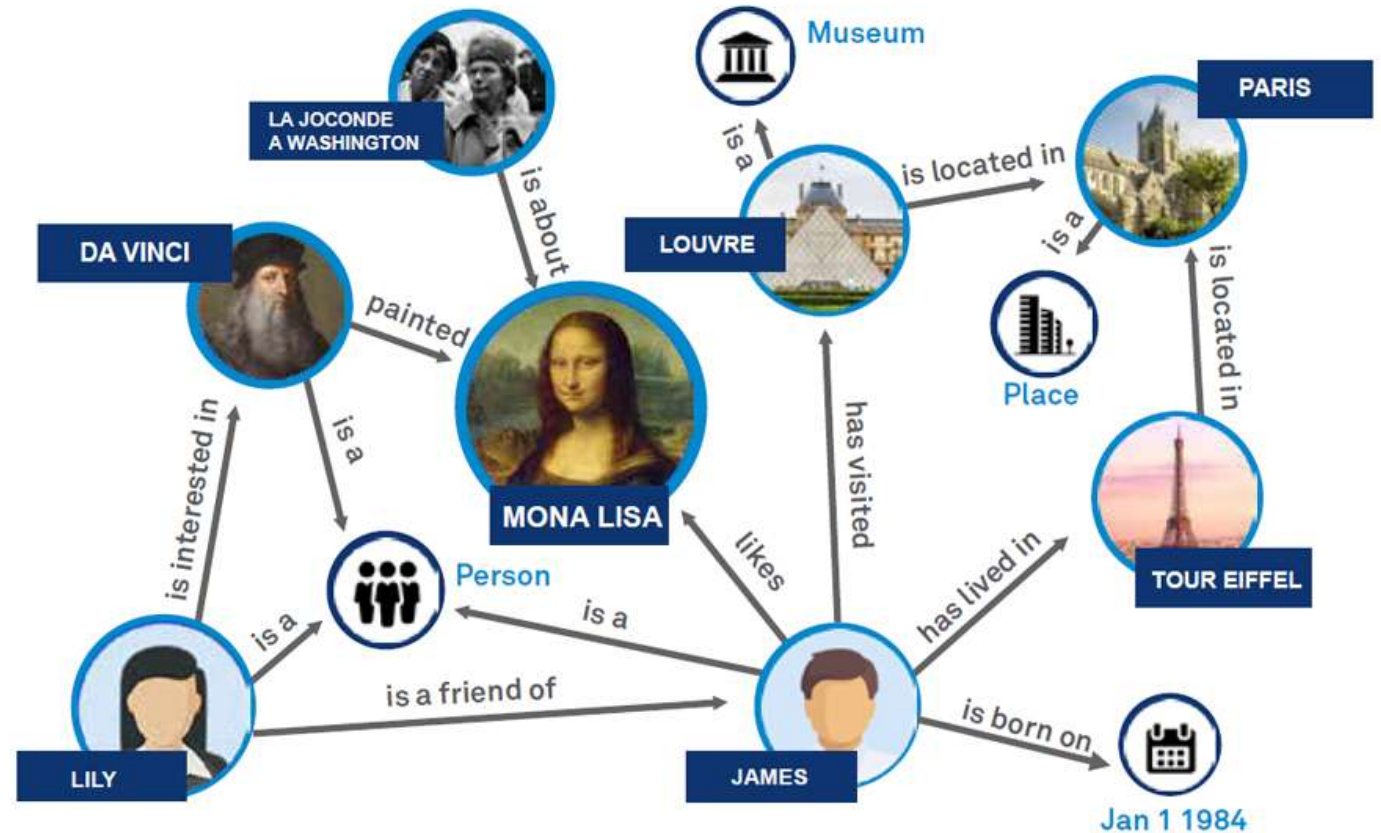
If X is a grandfather of Y then Y is a grandson of X

Exists X : `grandson(X,john)` ? Find a grandson of John



Knowledge graphs

A drastically **simplified form** of rule-based / expert systems



Solve a predicate logic reasoning task:

[help](#)

% a simple example: using two facts and two rules, find a grandson of john

```
father(john,pete).  
father(pete,mark).  
(father(X,Y) & father(Y,Z)) => grandfather(X,Z).  
grandfather(X,Y) => grandson(Y,X).  
  
grandson(X,john) => $ans(X).
```

Prove

simple

examples



complex

examples



or

Browse

Advanced

Clear result

Clear input



Result:

result: proof found

answer: \$ans(mark).

proof:

- 1: [in] $\neg \text{father}(X,Y) \mid \neg \text{father}(Z,X) \mid \text{grandfather}(Z,Y).$
- 2: [in] $\text{father}(\text{pete},\text{mark}).$
- 3: [mp, 1, 2] $\neg \text{father}(X,\text{pete}) \mid \text{grandfather}(X,\text{mark}).$
- 4: [in] $\text{father}(\text{john},\text{pete}).$
- 5: [mp, 3, 4] $\text{grandfather}(\text{john},\text{mark}).$
- 6: [in] $\neg \text{grandfather}(X,Y) \mid \text{grandson}(Y,X).$
- 7: [mp, 5, 6] $\text{grandson}(\text{mark},\text{john}).$
- 8: [in] $\neg \text{grandson}(X,\text{john}) \mid \$\text{ans}(X).$
- 9: [mp, 7, 8] $$$\text{ans}(\text{mark}).$

Masinõpe

- Lihtsustatult öeldes: “äärmuslikult vinge statistika”.
- Kõige tuntum meetodite klass on **närvivõrgud**, mis omakorda jagunevad paljudeks alamtüüpideks.
- Kasutusel ka suur hulk **mitte-närvivõrkude**-põhiseid masinõppe algoritme: vektormasinad, lähimad naabrid, juhustemetsad, ...



Machine learning: what is the point?



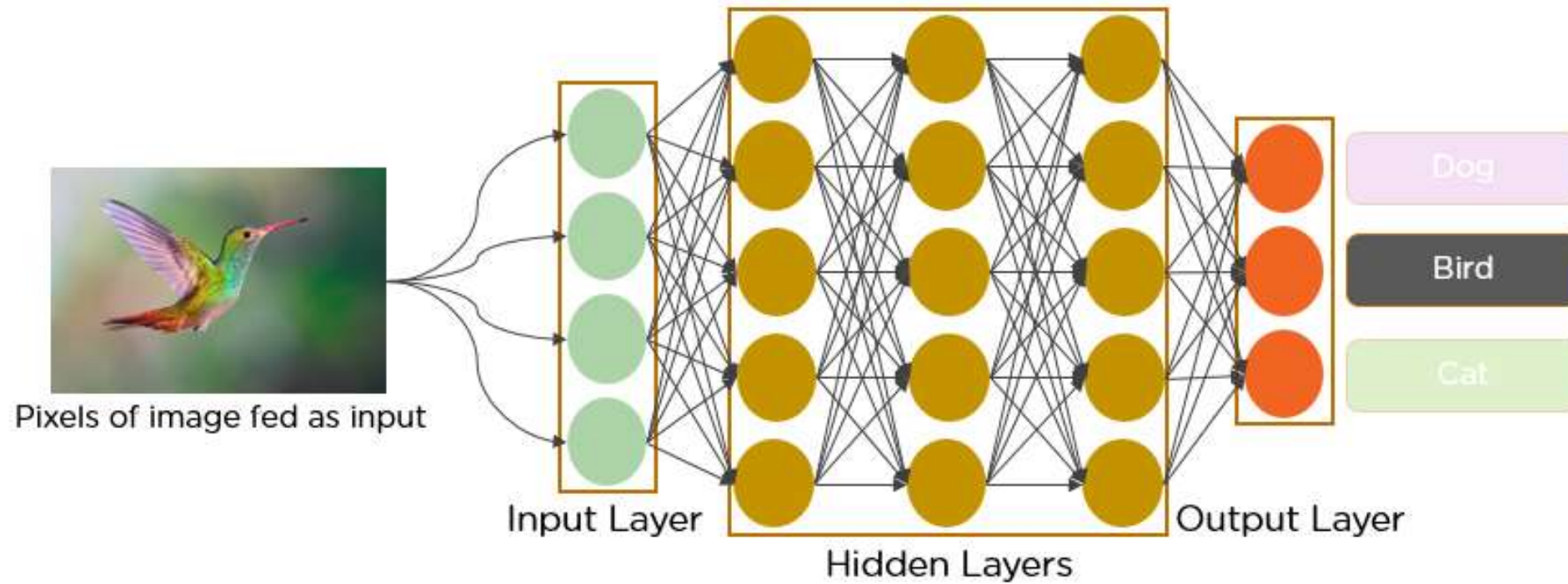
“cat”



“dog”

Writing a program to recognize whether an image is of a cat or a dog is theoretically possible, but extremely complicated: **not realistic in practice.**

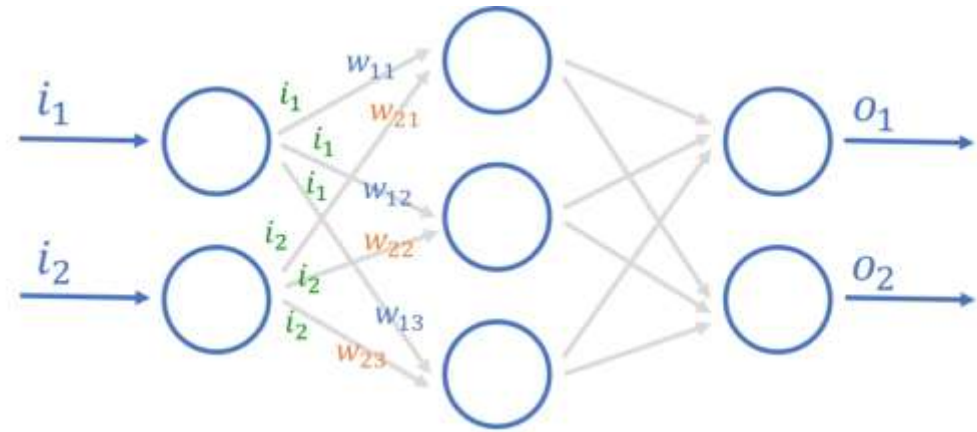
Deep neural networks ehk sügavad närvivõrgud



Matrices and calculations

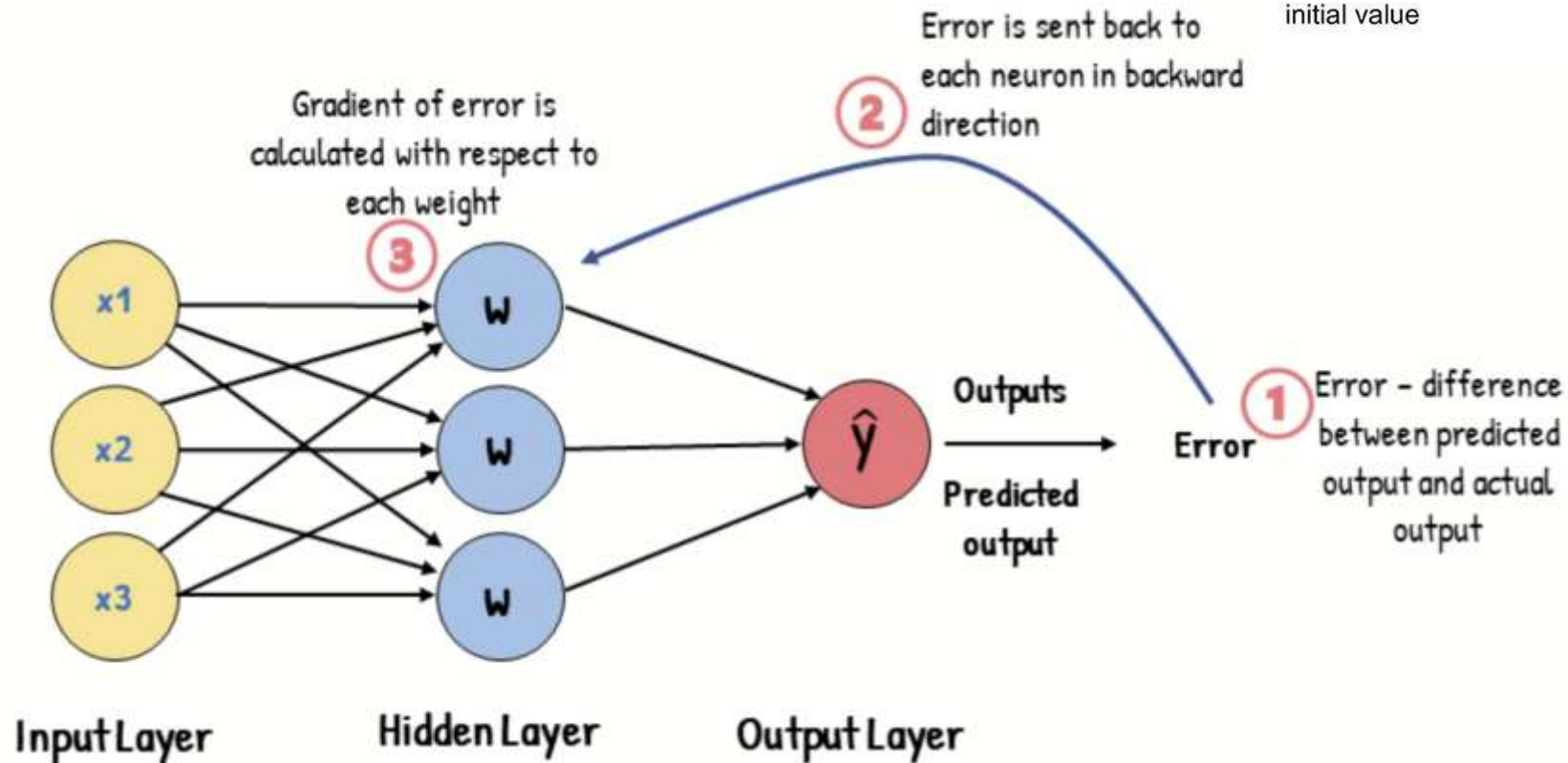
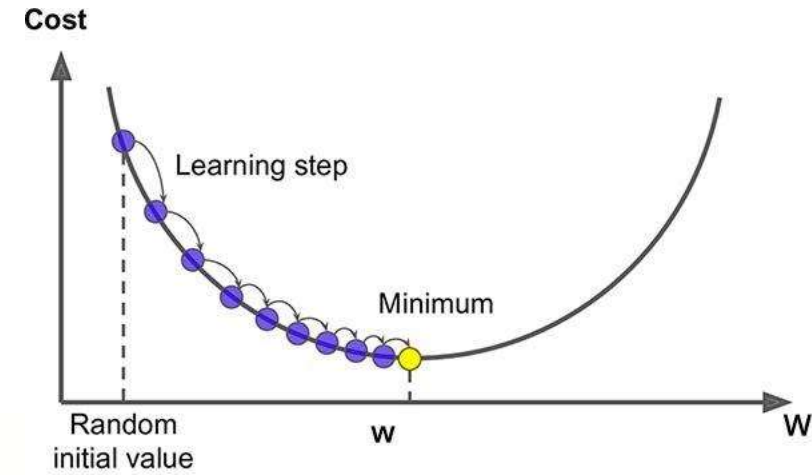
Connections are simply numbers (weights) you can store in a large table (matrix).

Calculating values for nodes can be done in several ways. One simple approach is to first multiply weights with inputs, sum all of them together and then normalize, using some popular simple function for this.

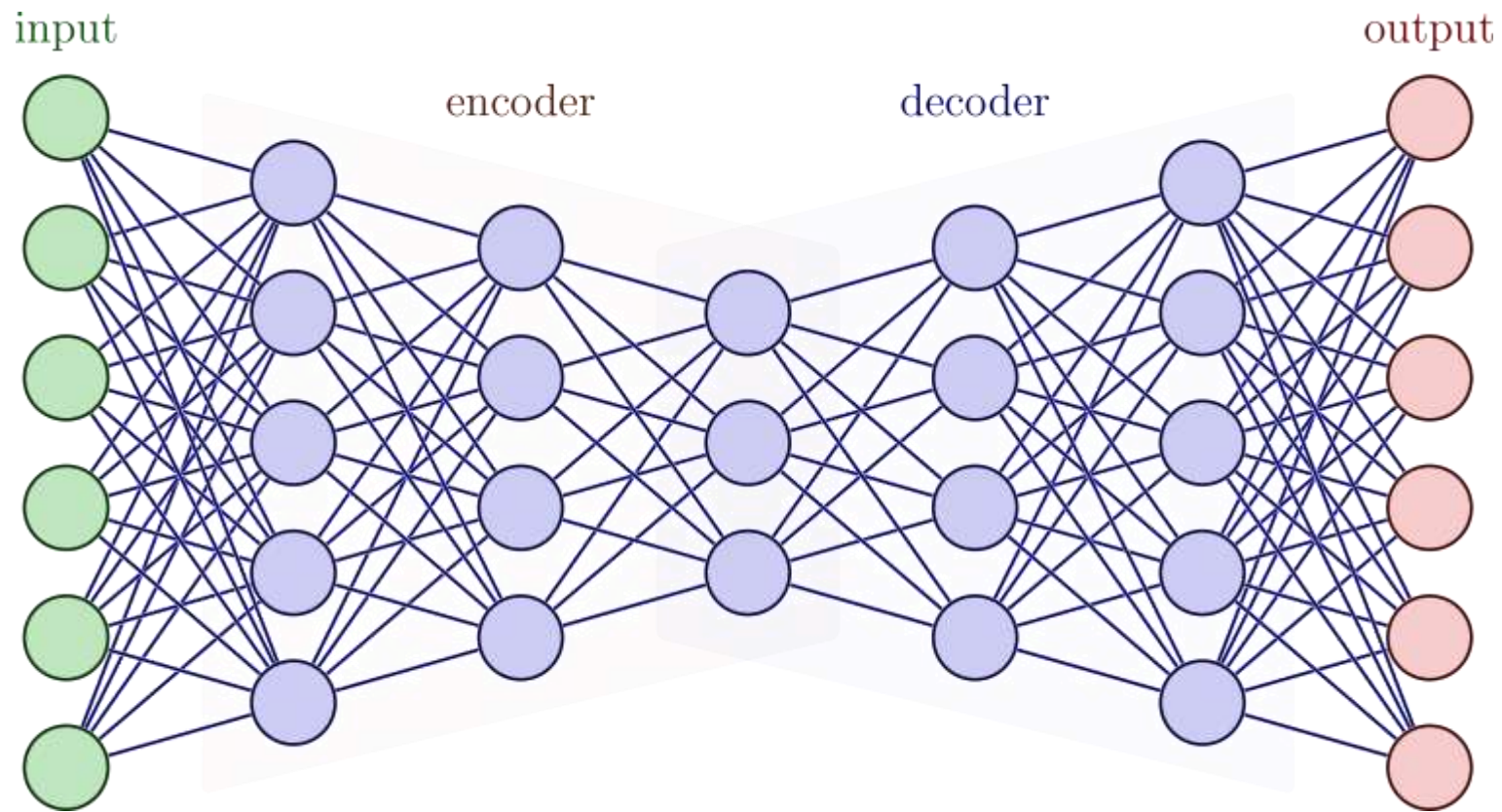


$$\begin{bmatrix} w_{11} & w_{21} \\ w_{12} & w_{22} \\ w_{13} & w_{23} \end{bmatrix} \cdot \begin{bmatrix} i_1 \\ i_2 \end{bmatrix} = \begin{bmatrix} (w_{11} \times i_1) + (w_{21} \times i_2) \\ (w_{12} \times i_1) + (w_{22} \times i_2) \\ (w_{13} \times i_1) + (w_{23} \times i_2) \end{bmatrix}$$

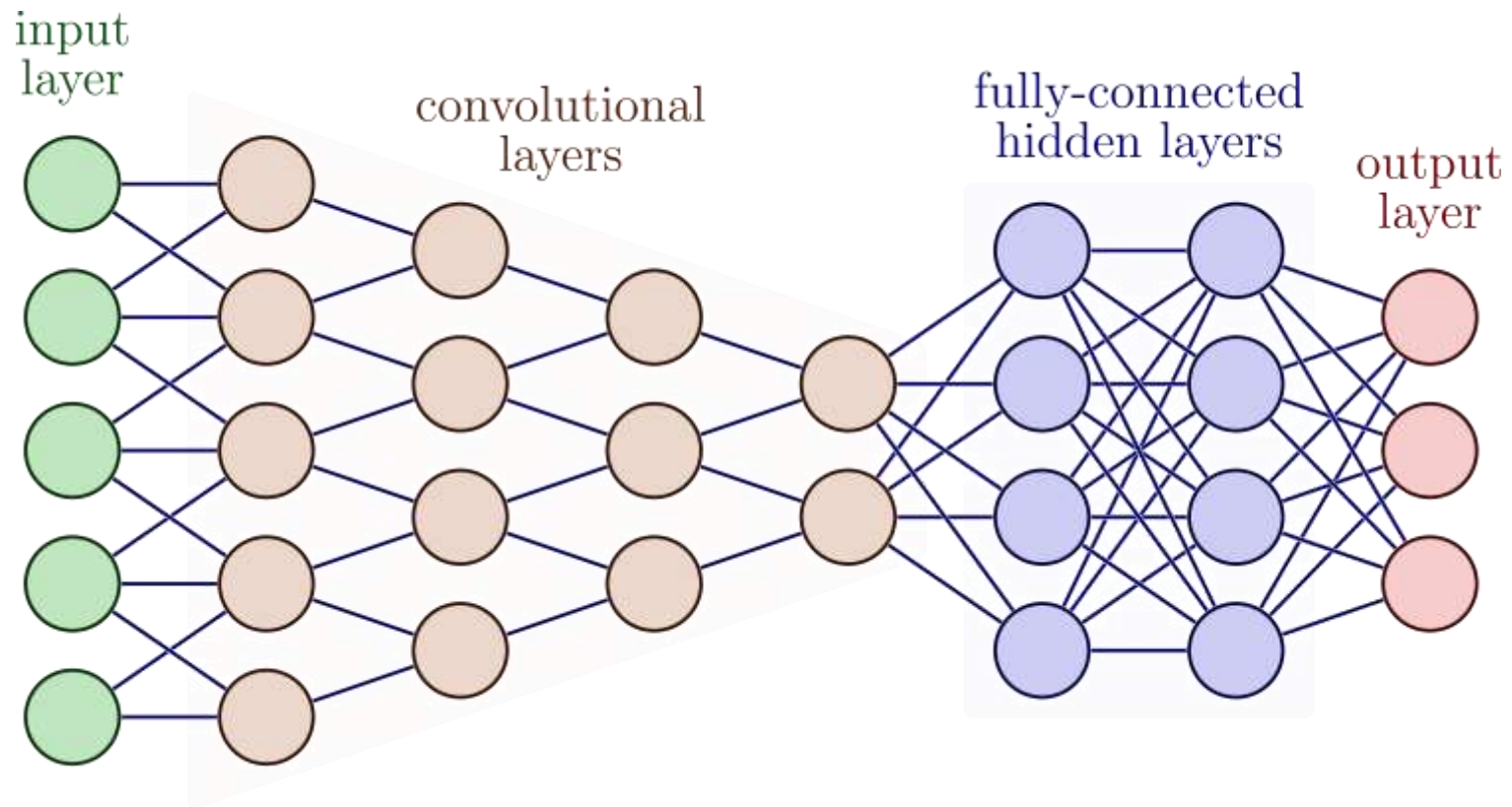
Learning by backpropagation and gradient descent



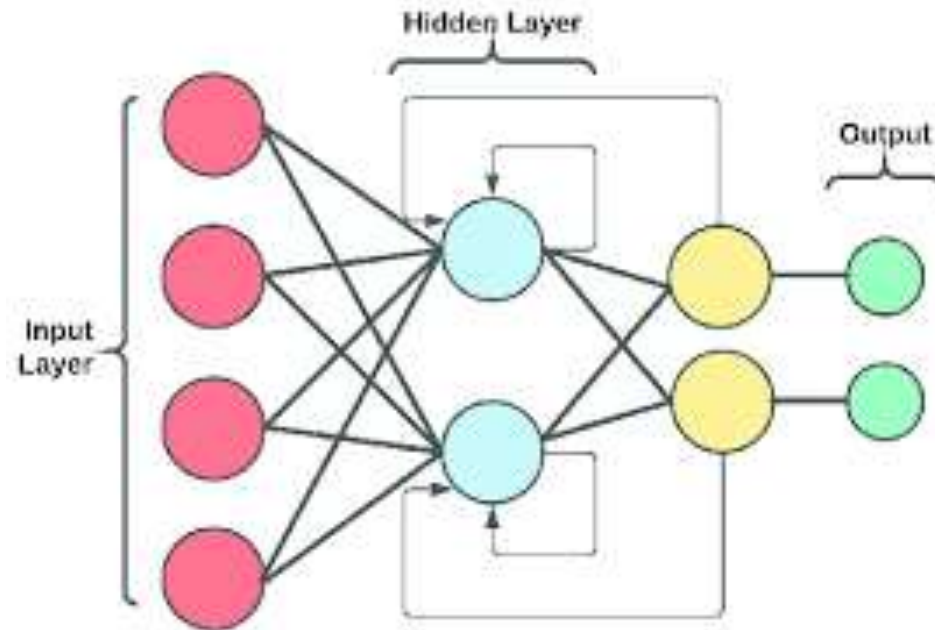
Autoencoder neural network (denoising, compression, generation)



Convolutional neural network (image recognition)



Recurrent neural network (time series, etc)



(a)



school bus 1.0

(b)



garbage truck 0.99

(c)



punching bag 1.0

(d)



snowplow 0.92



motor scooter 0.99



parachute 1.0



bobsled 1.0



parachute 0.54



fire truck 0.99



school bus 0.98



fireboat 0.98



bobsled 0.79

Mida suured keelemudelid (LLM) teevad?

Ennustavad samm-sammult teksti jätkumist:

- Jaan istus autosse ja ...
- Jaan istus autosse ja hakkas ...
- Jaan istus autosse ja hakkas sõitma ...
- Jaan istus autosse ja hakkas sõitma kodu ...

Kuidas LLM ennustab? Lihtsustatult:

- Kogub statistikat suure tekstihulga pealt: mismoodi seal on laused jätkunud, näiteks

→ Mihkel istus mootorrattale ja ... (*hakkas, mõtles, ...*)

- Kui näeb lauset, mille kohta statistikat ei ole, kasutab statistikat varasemate sarnaste lausete kohta:

→ Jaan istus autosse ja ...

LLM hallucinations: confidently generating wrong answers.

Why?

- LLMs learn an approximate, highly compressed picture of the world knowledge.
- LLMs are always hallucinating: just that mostly these correspond to reality
- LLM learning optimizes for confident tone of voice and pleasing answers
- ... etc ...

LLMi põhiprobleemid 2025 (chatbot age)

- Hallutsineerib ehk leiutab “fakte”, mis ei ole üldse tõesed.
- Vaikimisi omab ainult üldteadmisi.
- Ei ole mälu: uusi asju raske juurde õpetada.
- “Must kast” ehk ei anna tegelikke seletusi, kuidas vastused saadi.
- Ei sisalda otsialgoritme ja enamasti ei suuda teha pikemaid järeldusahelaid.
- Ei ole lihtsat viisi siduda andmebaasidega, inseneriarvutustega jne.
- Vastuste kvaliteet langeb kiirelt, kui sisendtekst on pikk.
- Ei ole selget viisi tema suunamiseks või mingit sorti vastuste keelamiseks.
- ...

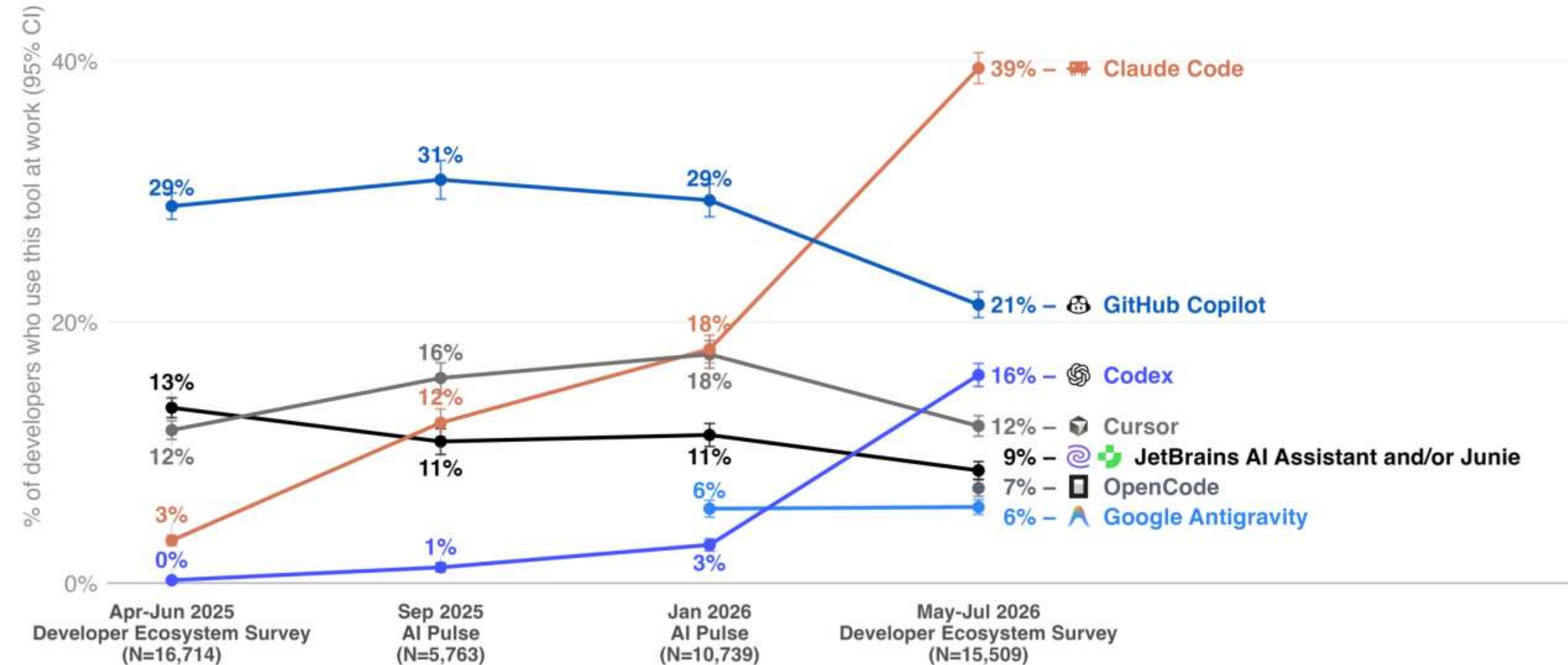
LLMi põhiprobleemid 2026 (agentic age) :

oranzhid probleemid püsivad, rohelised oluliselt paranenud, mustad kergelt paranenud

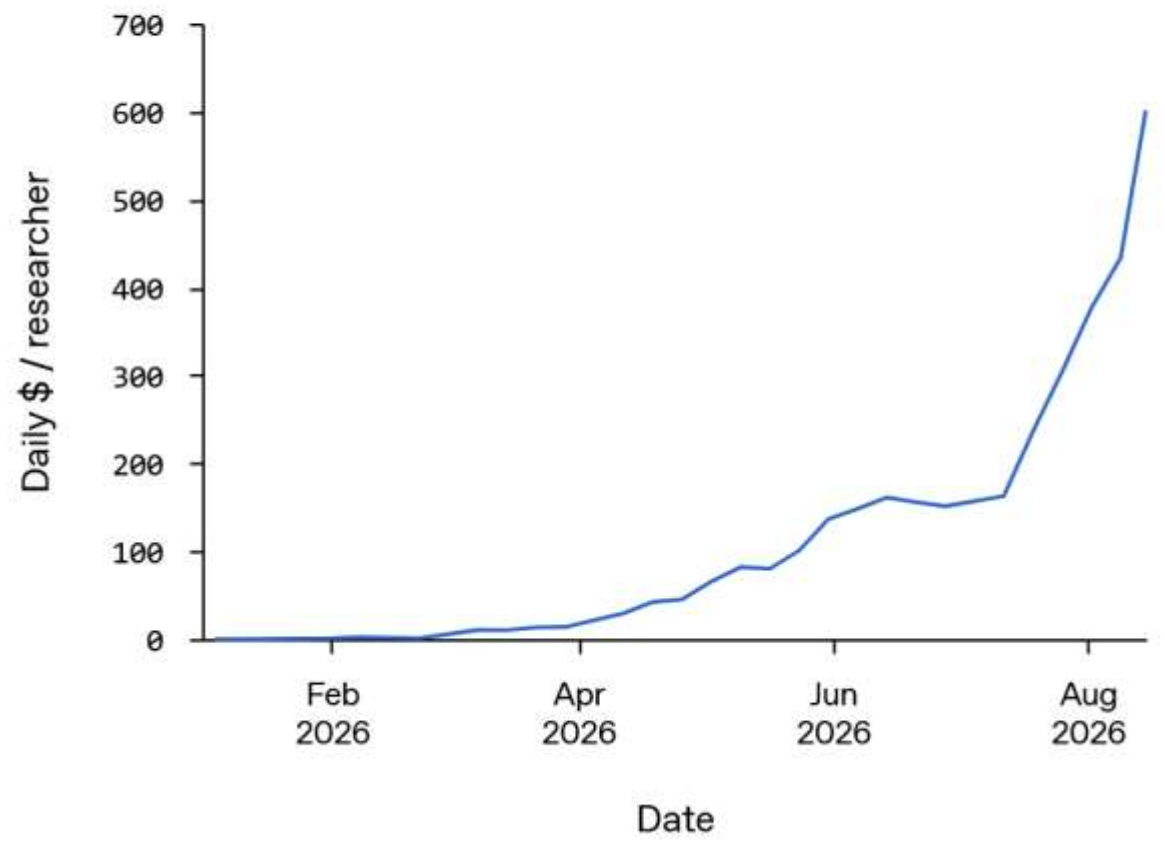
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- ...

Use at work

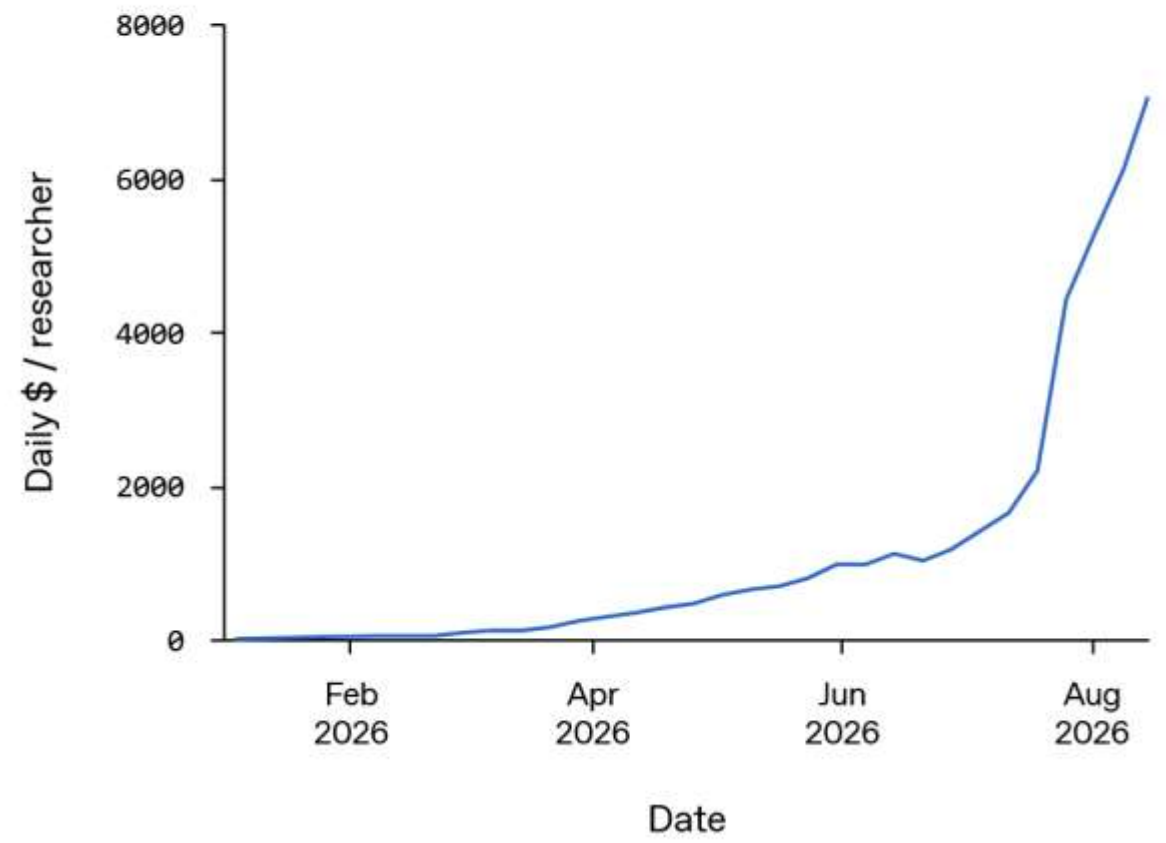
Which of these AI tools do you use for your coding and other development-related tasks at work?
(Multiple answers were possible; only a subset of tools is shown)



Usage of internal coding agents is increasing significantly—Median researcher



Usage of internal coding agents is increasing significantly—90th percentile researcher



Neurosümboolsed
ehk hübriidsüsteemid

	Common AI Techniques					
Use Case Families	Generative Models	Non-Generative ML	Optimisation	Simulation	Rules	Graphs
Forecasting	Low	High	Low	High	Medium	Low
Planning	Low	Low	High	Medium	Medium	High
Decision Intelligence	Low	Medium	High	High	High	Medium
Autonomous System	Low	Medium	High	Medium	Medium	Low
Segmentation	Medium	High	Low	Low	High	High
Recommender	Medium	High	Medium	Low	Medium	High
Perception	Medium	High	Low	Low	Low	Low
Intelligent Automation	Medium	High	Low	Low	High	Medium
Anomaly Detection	Medium	High	Low	Medium	Medium	High
Content Generation	High	Low	Low	High	Low	Low
Chatbots	High	High	Low	Low	Medium	High
Knowledge Discovery	High	Medium	Low	Low	Medium	High

Research area of neurosymbolic A.I. aka hybrid A.I.

Find ways to combine M.L. with symbolic components to be able to solve different really hard tasks.

Some approaches:

- Use M.L. to heuristically guide proof search for symbolic reasoning
- Build LLMs capable of autoformalization and then use symbolic reasoning
- Let LLMs learn on symbolic knowledge in addition to free text
- Build new architectures for M.L. capable of learning external software tool use
- Build new architectures for M.L. capable of performing search / planning
- Build symbolic rules already into a neural switch level of M.L. systems

Explainable A.I. ehk selgitatav tehismõistus

Masinõppe süsteemide tehtud otsuste taga on suur hägune võrgustik seoseid, mitte inim-arusaadavad reeglid.

Otsuste põhjendamiseks oleks lisaks hägusele võrgustikule vaja ehitada mõistetavad reeglid, mille abil otsust põhjendada.

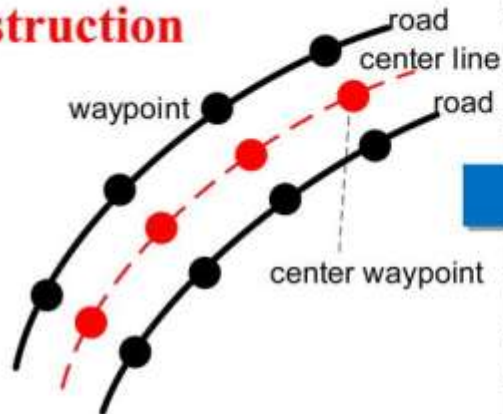
Hetkel puhtalt akadeemiline uurimistöö, praktilisi süsteeme ei osata teha.

*Hübriidsüsteemi näide:
isesõitvad autod*

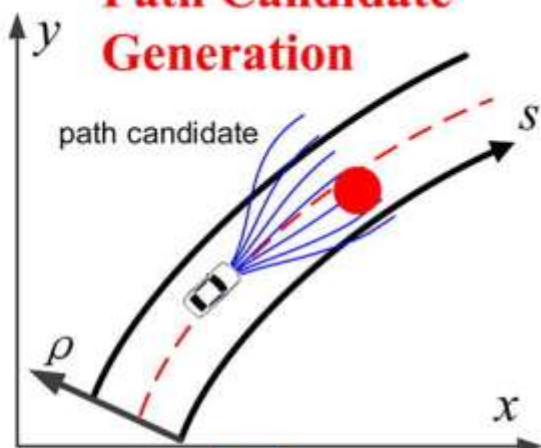
Pildil: 2005,
Stanfordi Stanley võidab
DARPA Grand Challenge II



Center Line Construction

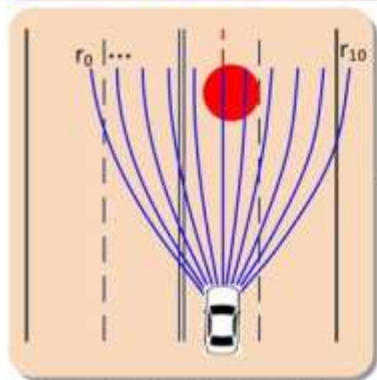


Path Candidate Generation

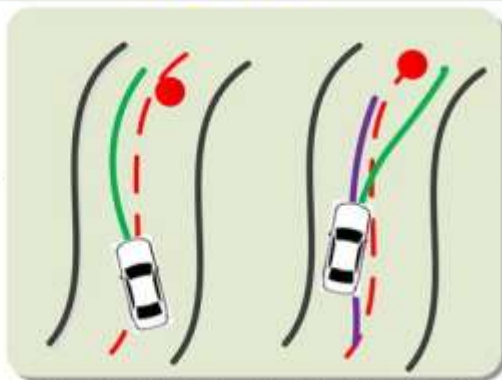


Path Selection

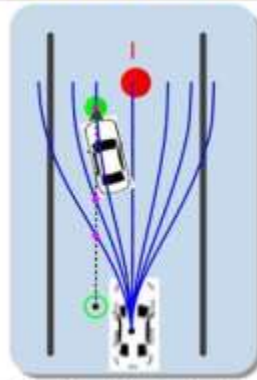
$$f = \omega_s f_s(r_i) + \omega_c f_c(r_i) + \omega_d f_d(r_i, a(r_i))$$



+

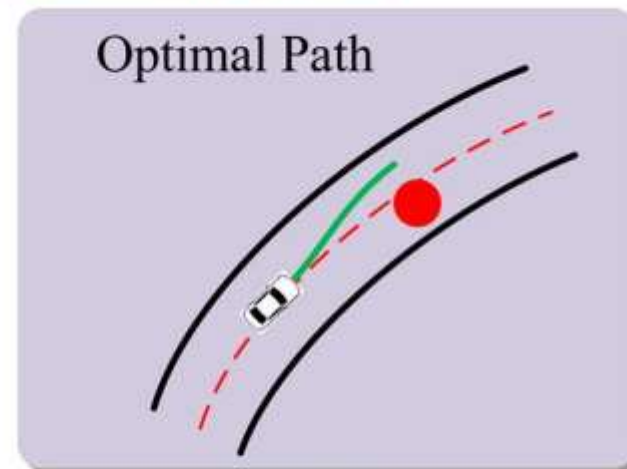


+

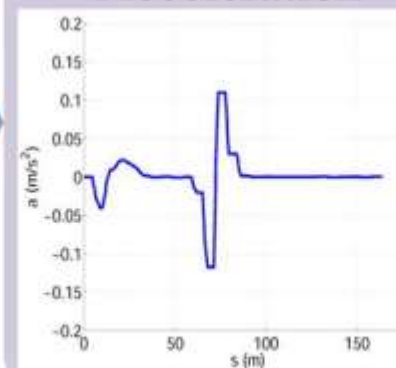


Path Planning Results

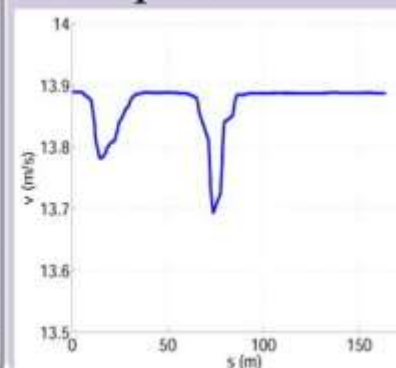
Optimal Path



Appropriate Acceleration



Appropriate Speed





WHEN CHILDREN ARE PRESENT

Avenue

38
TERRACE
Underpass & Station

Rotini

cruise

May stop quickly

23810T3

Rotini

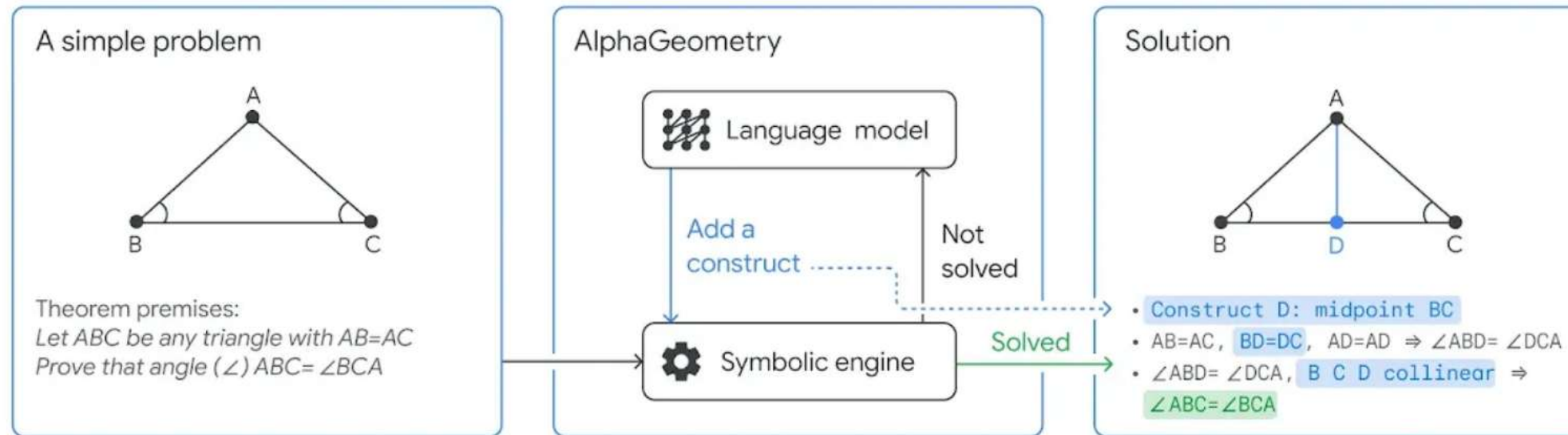
AlphaGo ja AlphaZero: hübriidsüsteemid

DeepMind / Google projekt masinõppeks Go ja male jaoks.

Iseenda vastu mängides õpib seisumustrite järgi seisu headust kiirelt ja hästi hindama.

Hea seisuhindamine on kombineeritud **Monte-Carlo otsinguga**: vaatab juhusliku valikuga läbi hulga võimalikke käikude jadasid mängu lõpuni välja.

Google DeepMind AlphaGeometry: autoformalization + symbolic reasoning engine



Solves complex geometry problems at a level approaching a high-school Olympiad gold-medalist

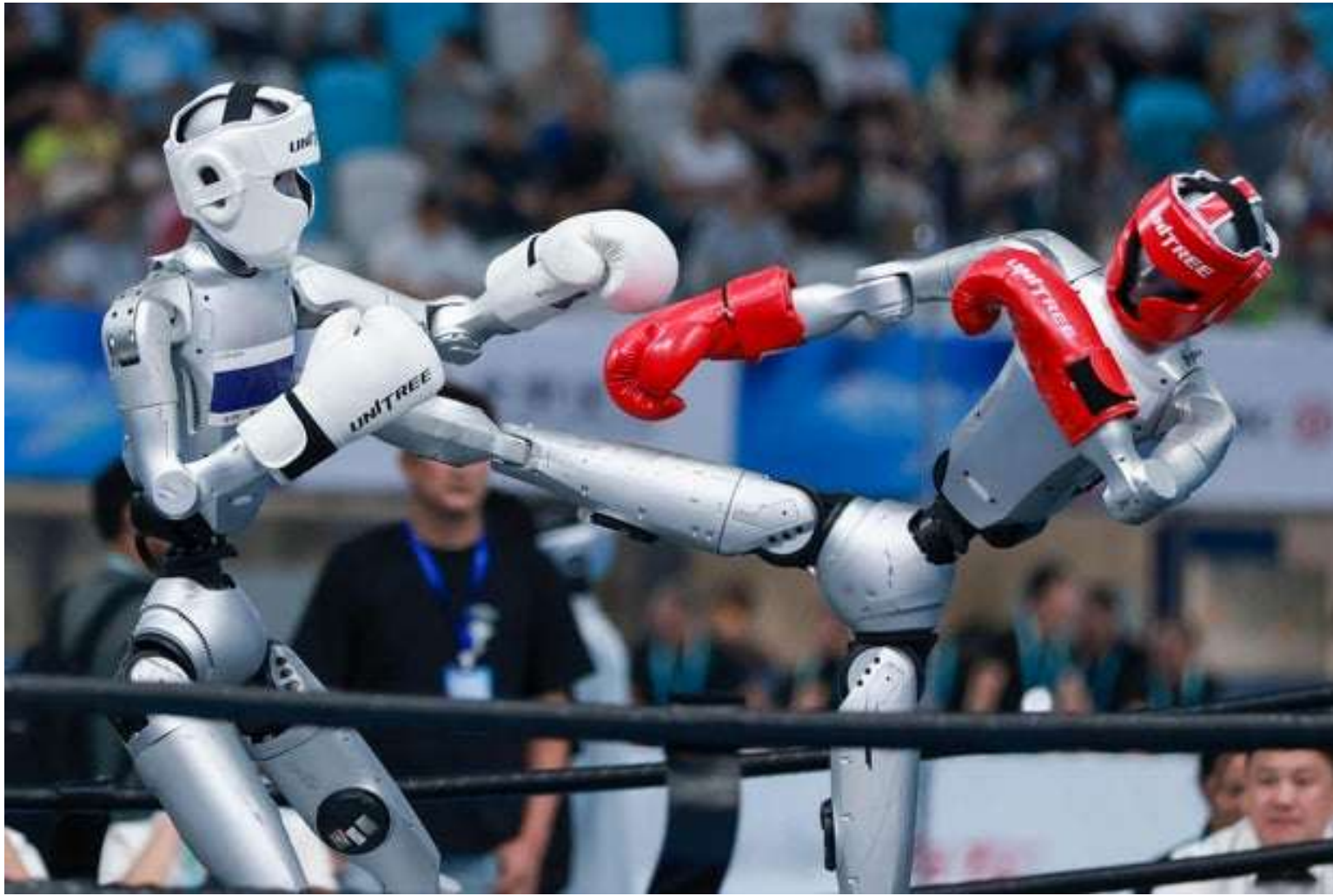
Robotid

*Robots: a **very** hard area!*



Most robots contain fairly little A.I.



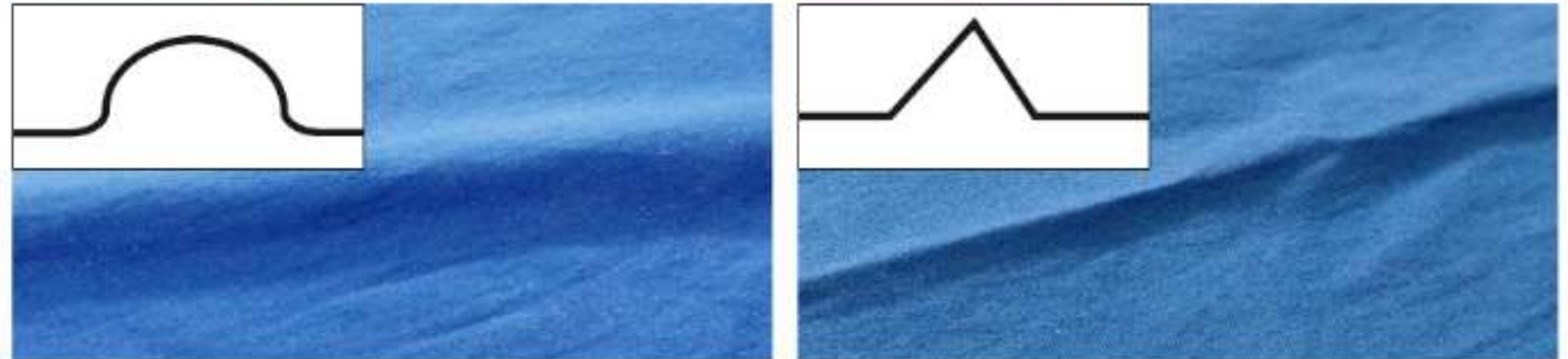


Industrial robots are **mostly hand-coded**,
with A.I. relegated to detection etc



Example:

No real success in developing robots able to iron ordinary clothes.



Example:

Bringing dirty dishes from the table to the dishwasher does not seem realistic in visible future.





Project: Dynamics / MIT CSAIL Robotics

cheddar

TikTok
@cheddar

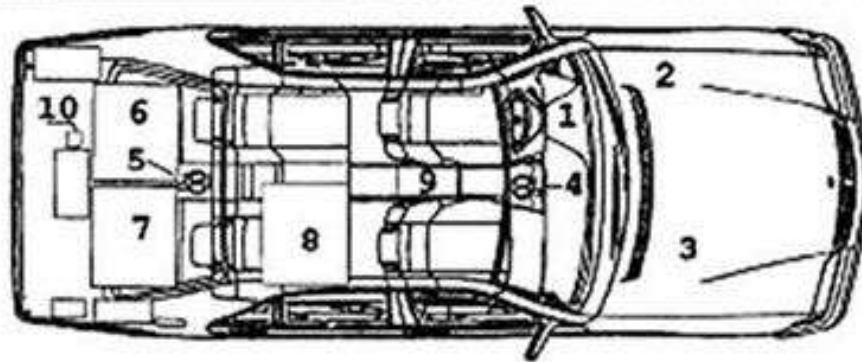
@sarahstacks_



TikTok
@espn

6.127 Second Cycle
Second Fastest Time in US

1980- ja 1990-d:
Isesõitev auto
Saksamaal:
tuhanded
kilomeetrid
kiirteel



- | | |
|-----------------------------|--|
| 1 Torque motor for steering | 6, 8 Transputer system, image processing |
| 2 brake system | 7 processors for gaze & locomotion control |
| 3 electric throttle control | 8 user interface |
| 4 front platform | 9 linear accelerometers |
| 5 rear platform | 10 angular rate sensors |

https://en.wikipedia.org/wiki/History_of_self-driving_cars

2005: Stanfordi
Stanley võidab
DARPA Grand
Challenge II



Darpa urban challenge 2007

96 km linnas, aeg 4:10



Tesla

„Traffic-Aware Cruise Control **cannot detect all objects and may not brake/decelerate for stationary vehicles**, especially in situations when you are driving over 50 mph (80 km/h) and a vehicle you are following moves out of your driving path and a stationary vehicle or object is in front of you instead.”

Volvo

„In other words, **your Volvo won't brake to avoid hitting a stopped car that suddenly appears up ahead**“.

Filosoofia ja prognoosid

*Tugev ehk „päris“ AI, mida meil ei ole (???)
(ehk AGI: artificial general intelligence):*

Masin on intellektilt inimloomusele lähedane või ületab seda, suudab täita tüüpilisi inimülesandeid, kasutada laia spektrit taustateadmisi ja omab mingil määral eneseteadvust.

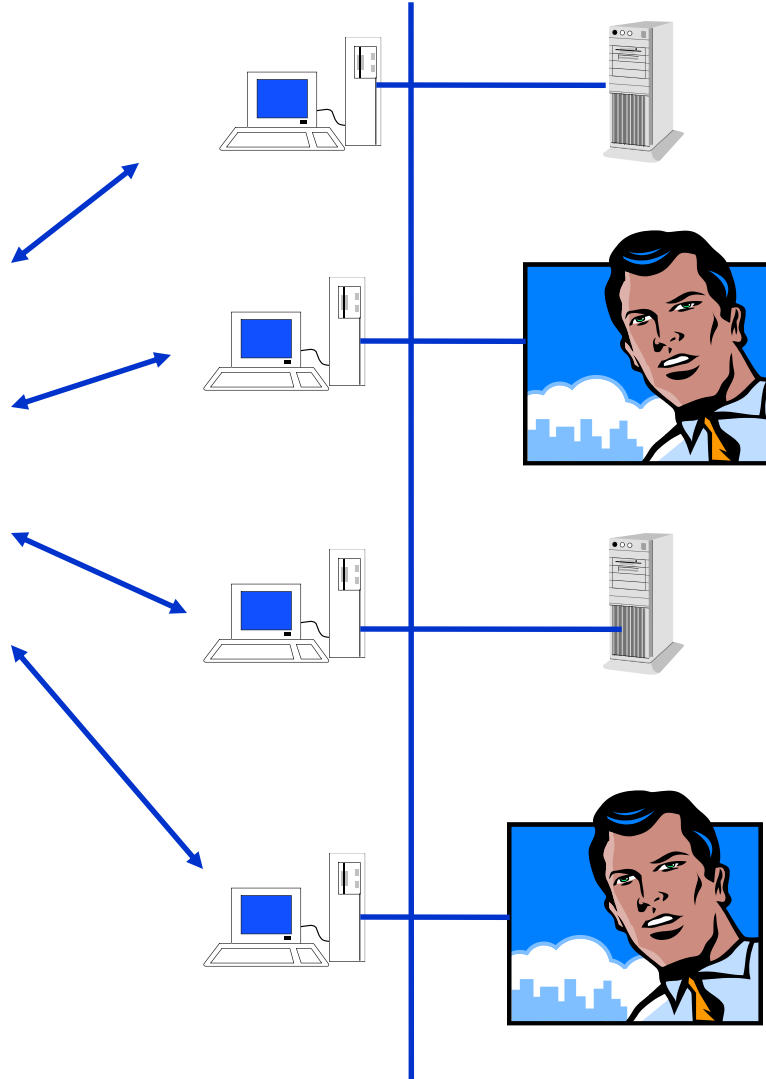
Nõrk AI ehk see, mis tegelikult on:

Tarkvara kasutamine spetsiifiliste probleemide või mõtlemisülesannete lahendamiseks, mis ei vaja laia spektrit inimvõimeid.

Turingi test: filosoofiline idee/ettepanek 50-datest



Mõistata, kas
chati-ekraani
taga on inimene
või programm?



Turing:

Kui katsetajad
ei suuda ära
arvata (st
ära-arvamise
sagedus on
50% ja 50%
eksitakse), siis
on jutlev
masin
päriselt
intelligentne.



**IF ANYONE
BUILDS IT,
EVERYONE
DIES**

**WHY
SUPERHUMAN AI
WOULD
KILL US ALL**

**ELIEZER
YUDKOWSKY &
NATE SOARES**

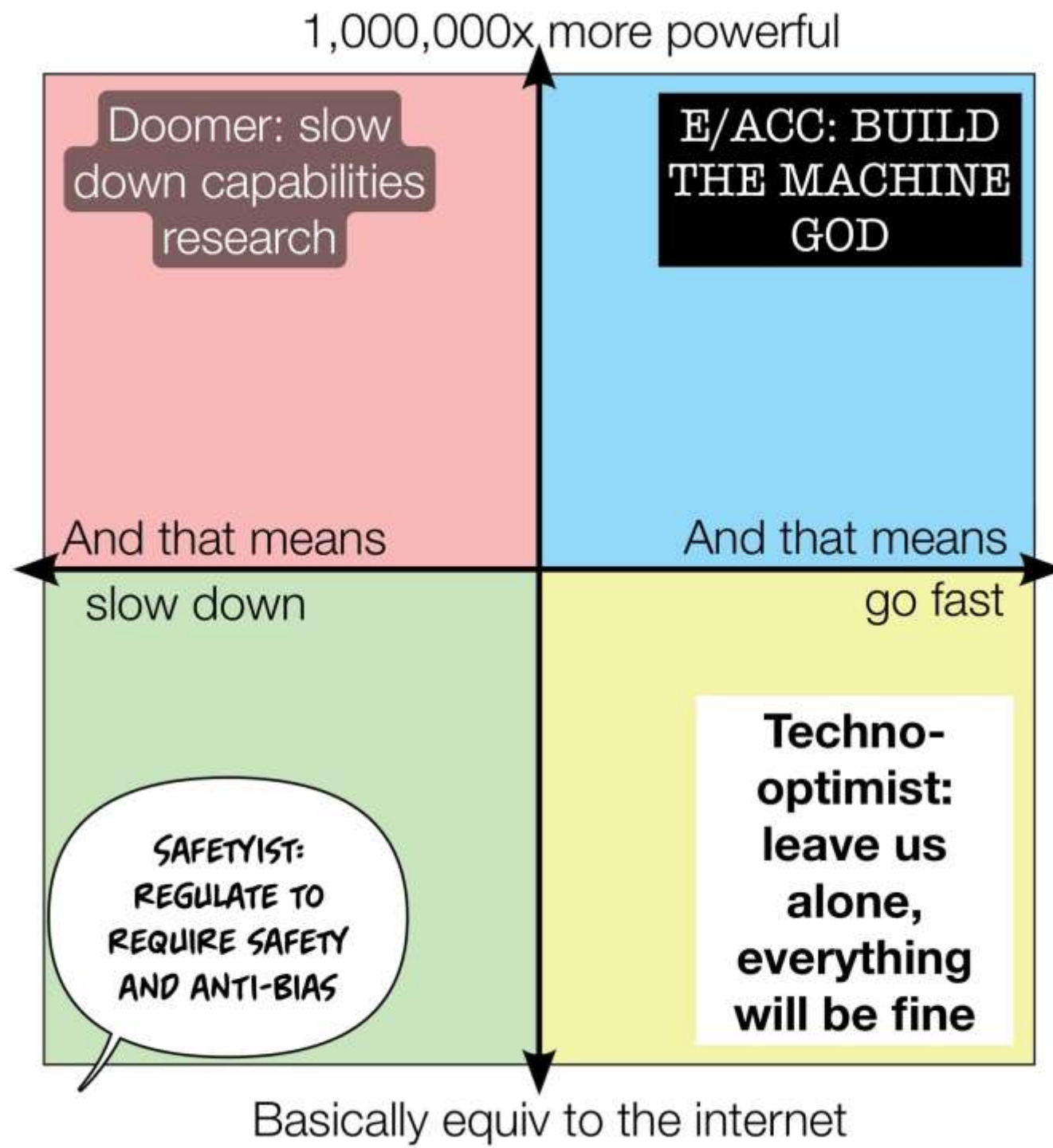
WHY AI WILL SAVE THE WORLD

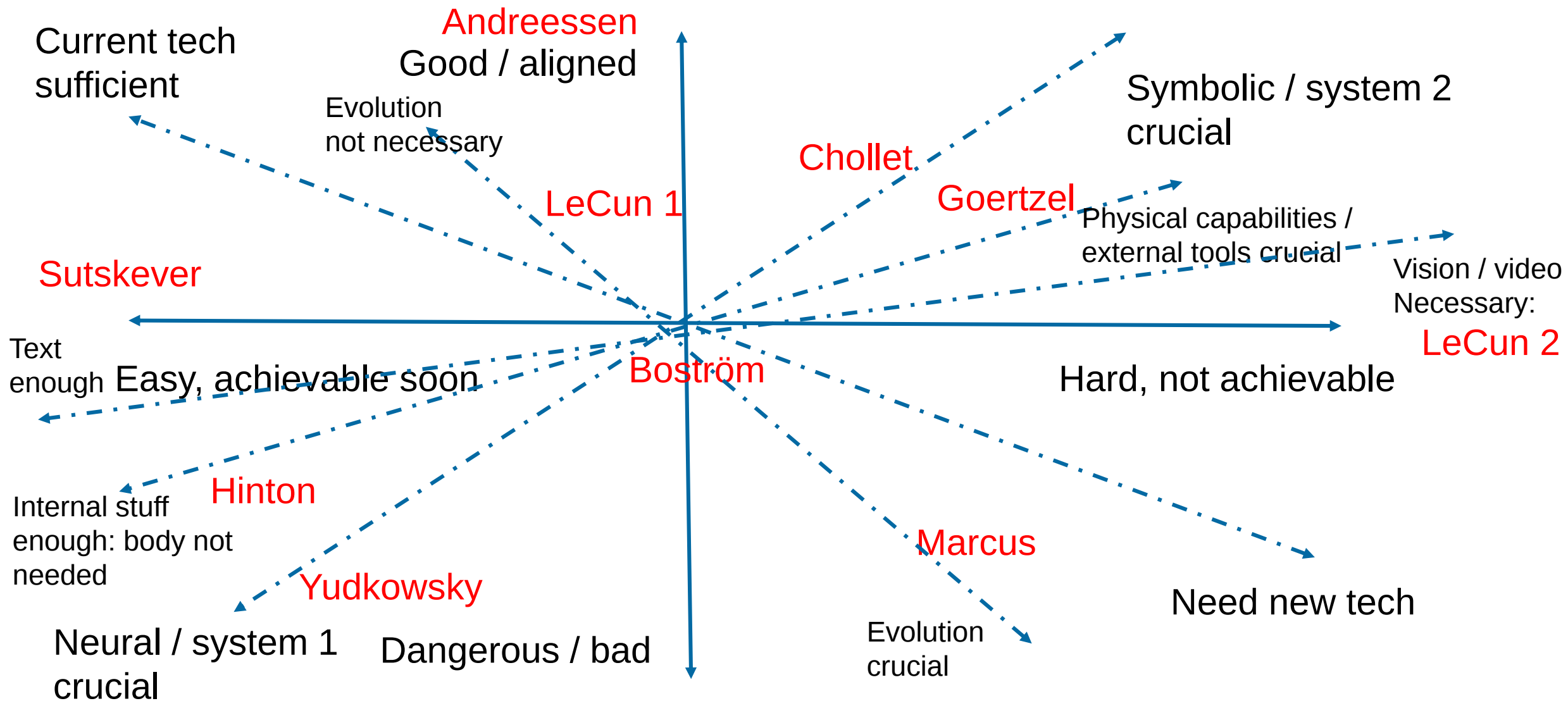
**MARC
ANDREESSEN**



The AI Race Will **Determine** How We All Live

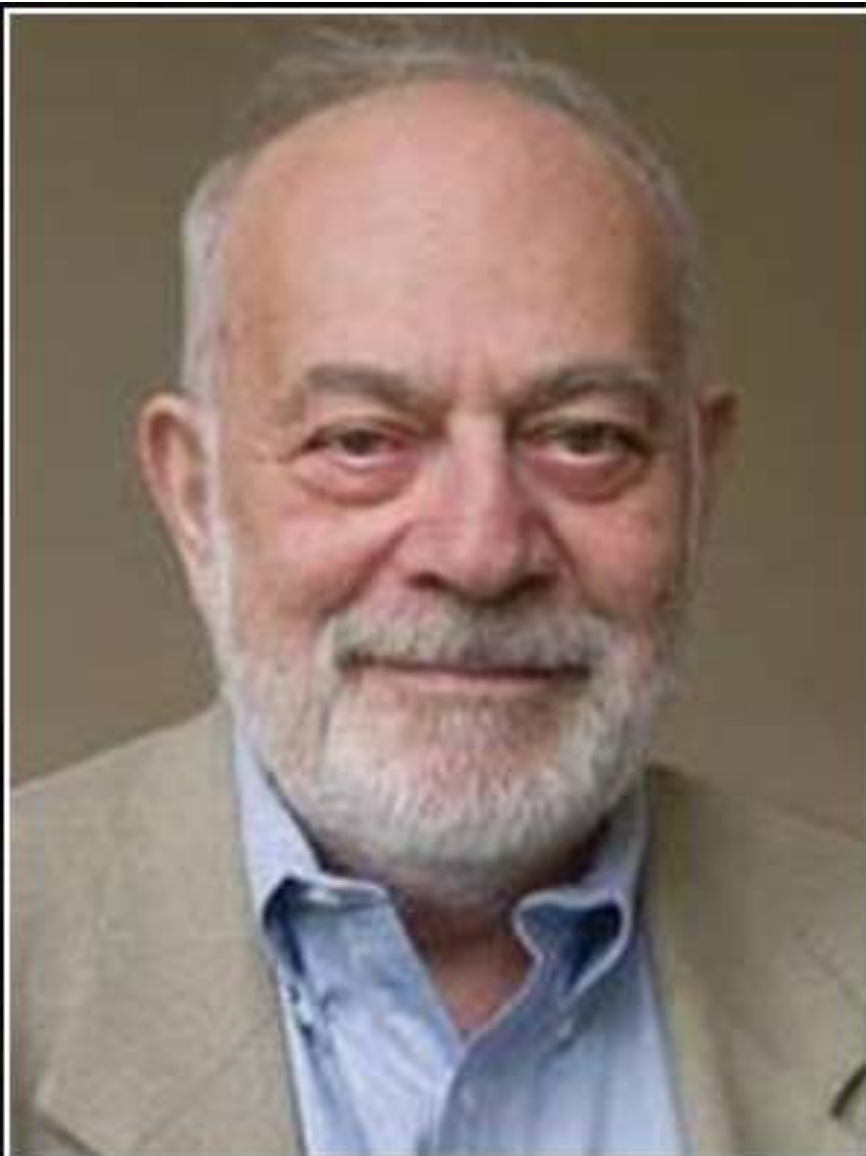






ON BULLSHIT

Harry G. Frankfurt



The bullshitter is neither on the side of the true or the side of the false. His eye is not on the facts at all. He does not reject the authority of the truth, as the liar does, and oppose himself to it. He pays no attention to it at all. By virtue of this, bullshit is a greater enemy of the truth than lies are.

— *Harry Frankfurt* —

AZ QUOTES



European Union A.I. Act: 12 July 2024

The Act assigns applications of AI to several risk categories.

Applications and systems that create an **unacceptable** risk are banned (manipulative / deceptive, exploiting vulnerabilities, inferring emotions in organisations, unrestricted collection of facial databases, social scoring of the type used in China, etc).

High-risk applications (profiling individuals, remote biometric identification systems, CV-scanning tool that ranks job applicants, determining access, admission or assignment to educational and vocational training institutions, applying the law to disputes, etc) are subject to specific legal requirements

Limited risk AI systems are subject to lighter transparency obligations: developers and deployers must ensure that end-users are aware that they are interacting with AI (chatbots and deepfakes).

Minimal risk is unregulated (including AI applications like AI enabled video games and spam filters).

Tasks AI Can NOT Do

Tasks AI Can Do

- Equal Difficulty Tasks
- - - Jagged Frontier
- Task On Frontier
- Task Outside Frontier
- Task Inside Frontier

Jagged Frontier of AI Task Capabilities

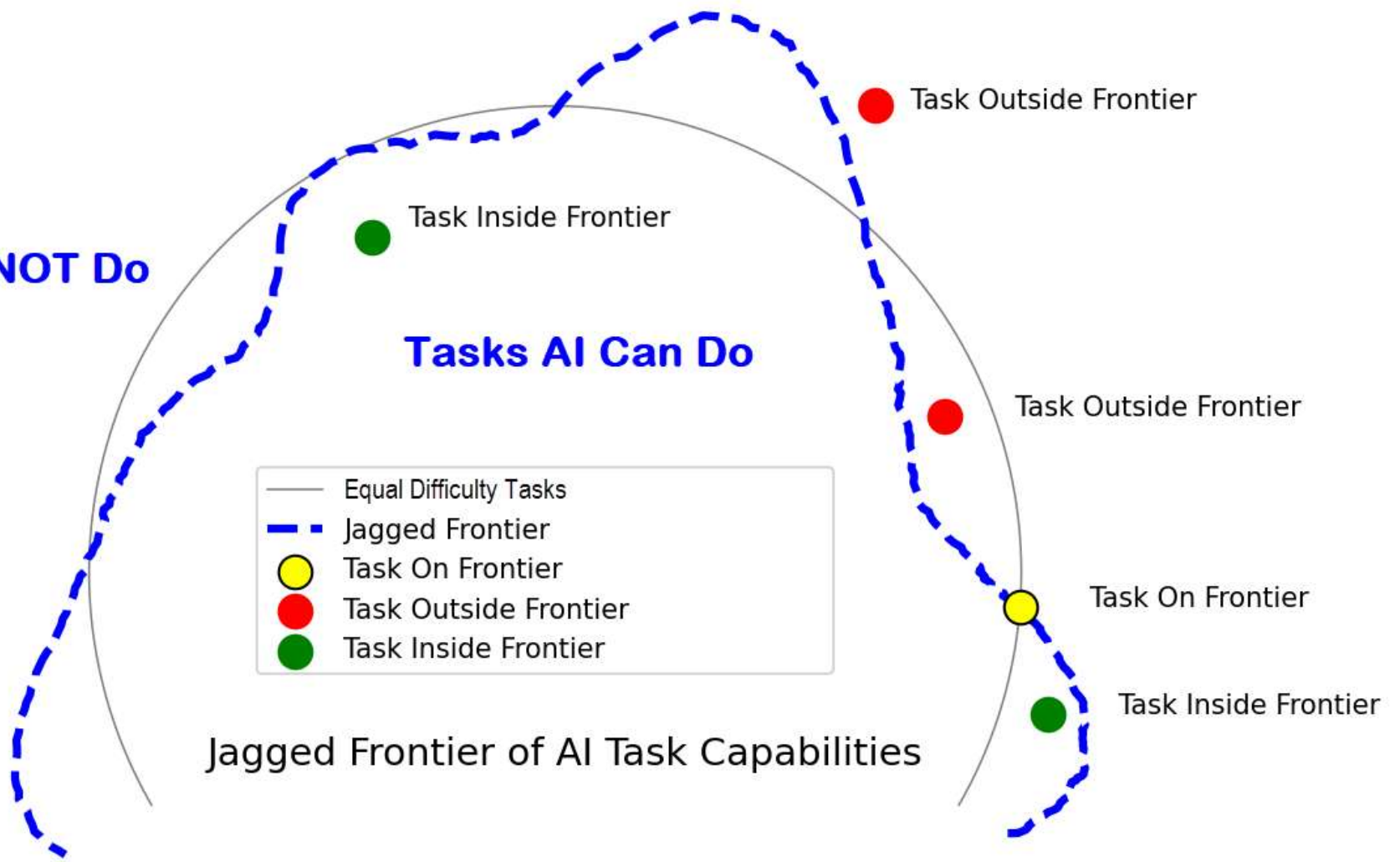
Task Outside Frontier

Task Inside Frontier

Task Outside Frontier

Task On Frontier

Task Inside Frontier



Moravec'i paradoks:

Inimeste jaoks rasked ülesanded on masina jaoks tihtipeale lihtsad.

Inimeste jaoks lihtsad ülesanded on masina jaoks tihtipeale äärmiselt keerulised.

Amara printsiip:

Meil on kalduvus lähiperspektiivis tehnoloogia mõju üle hinnata ja pikemas perspektiivis alahinnata.

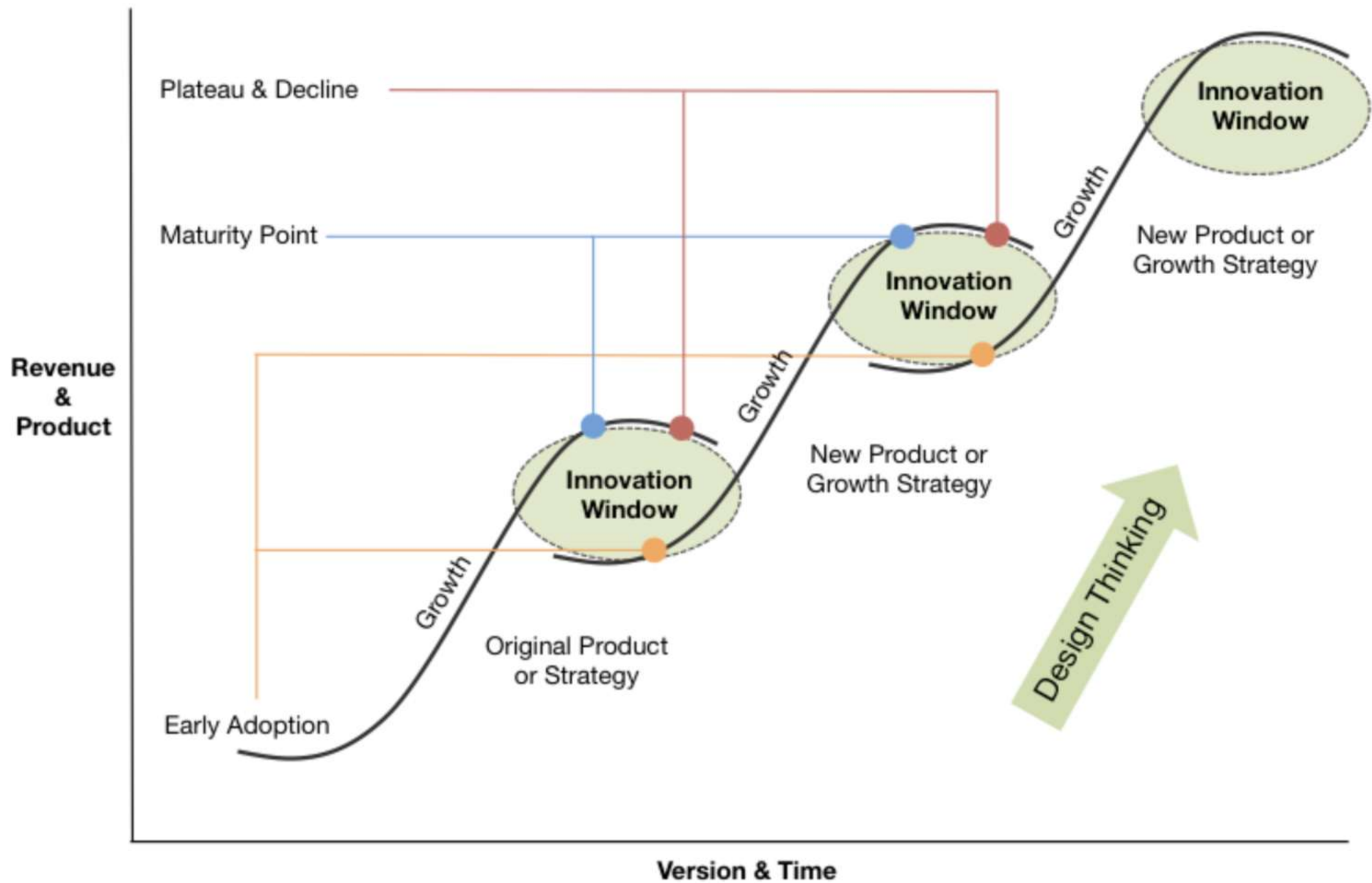
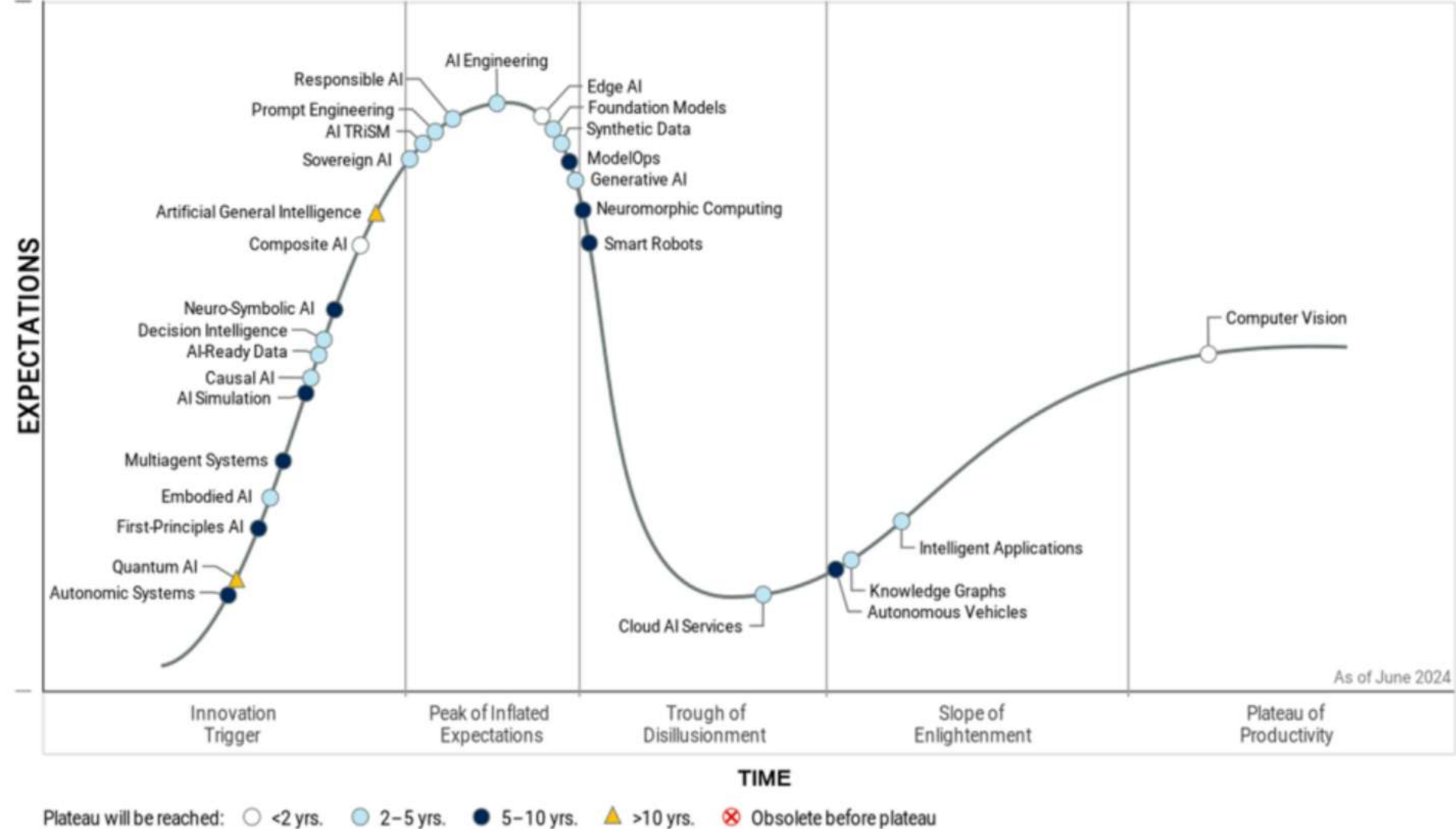
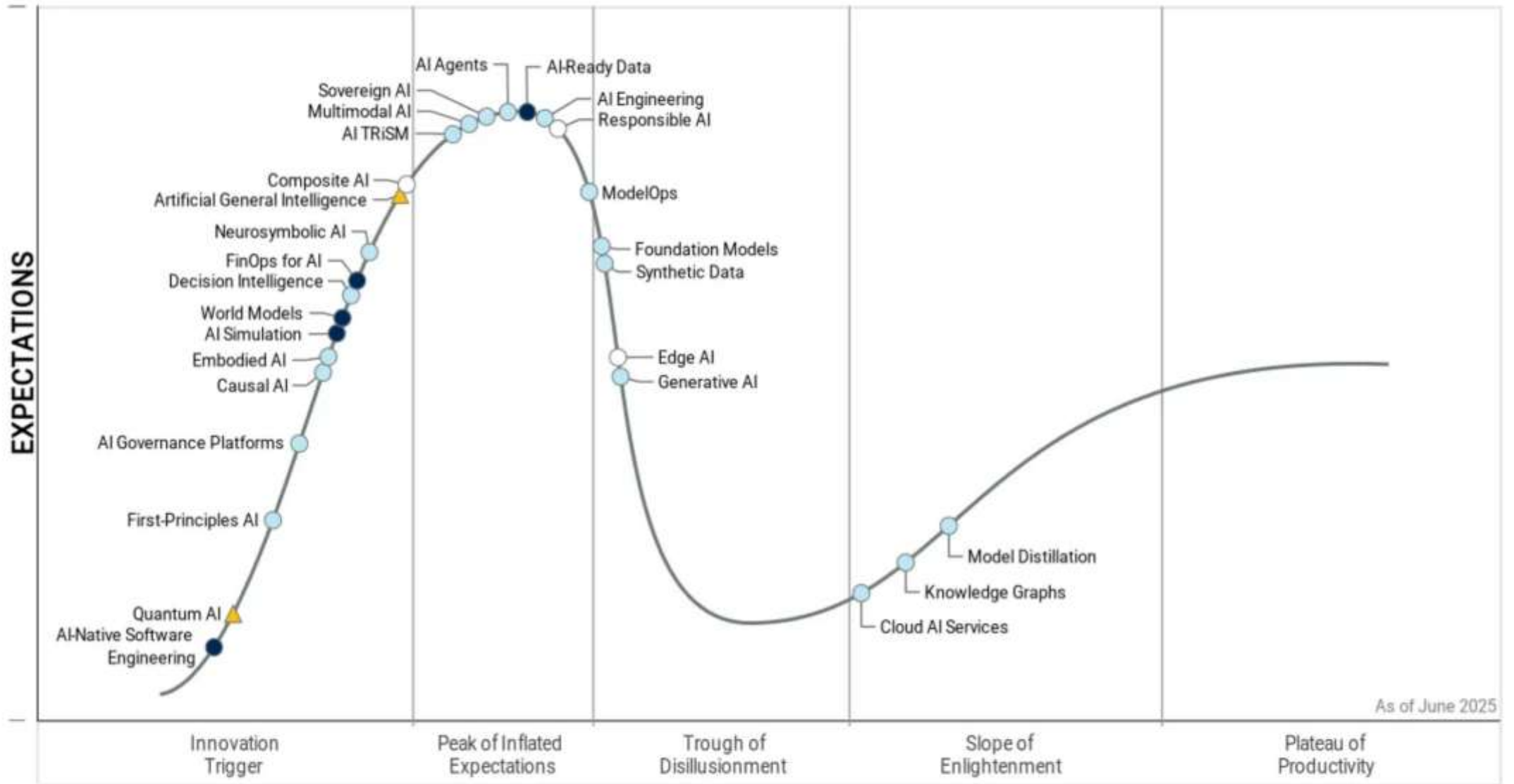


Figure 1: Hype Cycle for Artificial Intelligence, 2024

Hype Cycle for Artificial Intelligence, 2024

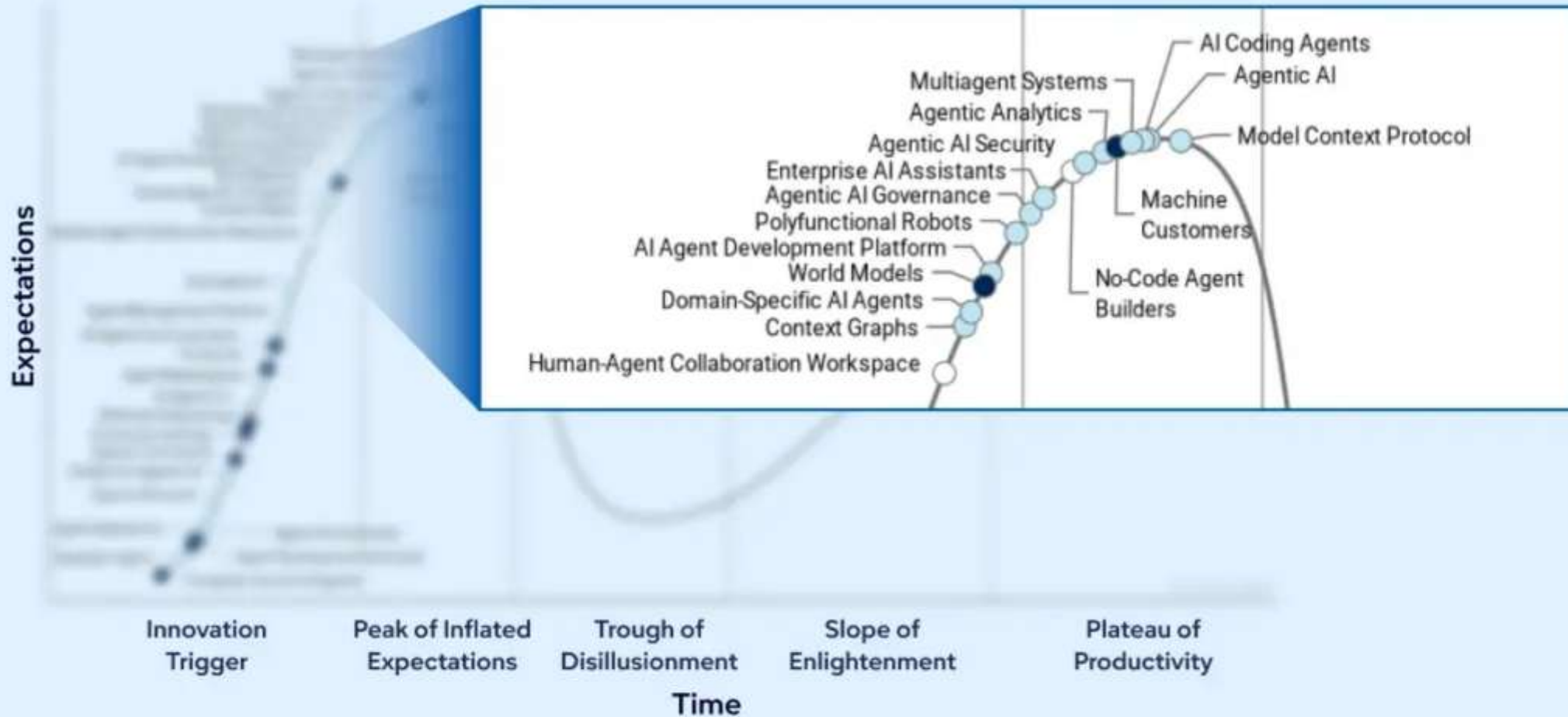


Hype Cycle for Artificial Intelligence, 2025



Hype Cycle for Agentic AI

Plateau will be reached ○ < 2 years ○ 2 – 5 years ● 5 – 10 years ▲ >10 years ⊗ Obsolete before plateau



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Gartner

Erinevad suunad, eesmärgid, tehnoloogiad:

- Tsentraliseerimine
- Automatiseerimine
- Tehisintellekt



Vastastikune mõju
ja võimendus

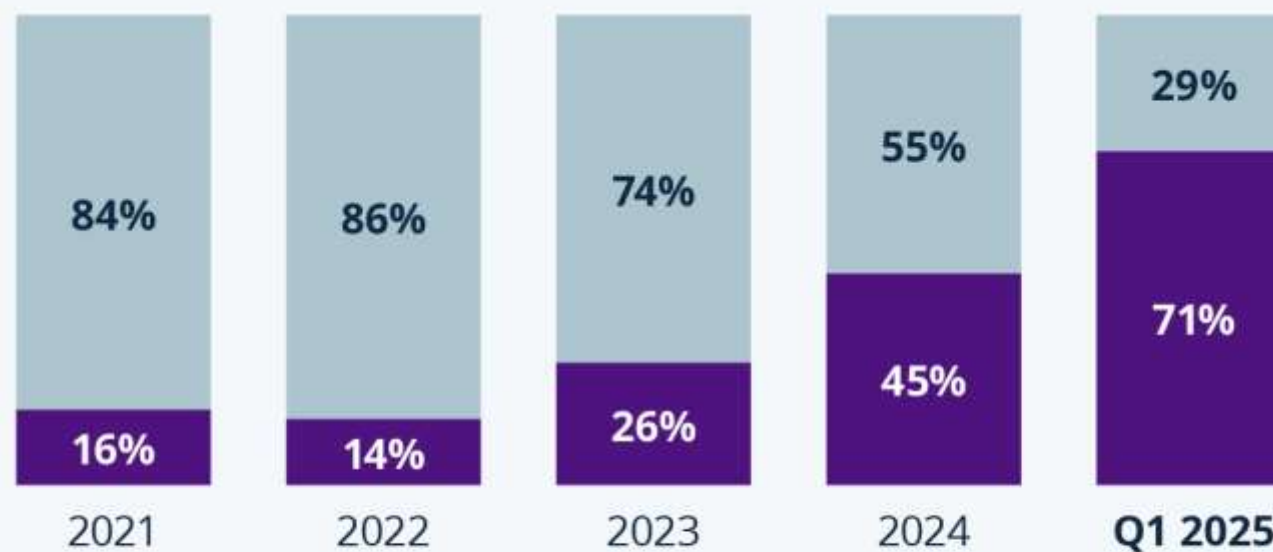
Keelemudelite äri ja kasutuse hinnangud
2025-2026

AI Sucks Up a Growing Chunk of VC Funding in the U.S.

Share of equity investments into U.S. VC-backed companies that are (not) AI-related*



■ AI-related investments ■ Non-AI investments



* Sources of cash investments include VC firms, corporate investors, other private equity firms and individuals.

Source: Crunchbase via EY



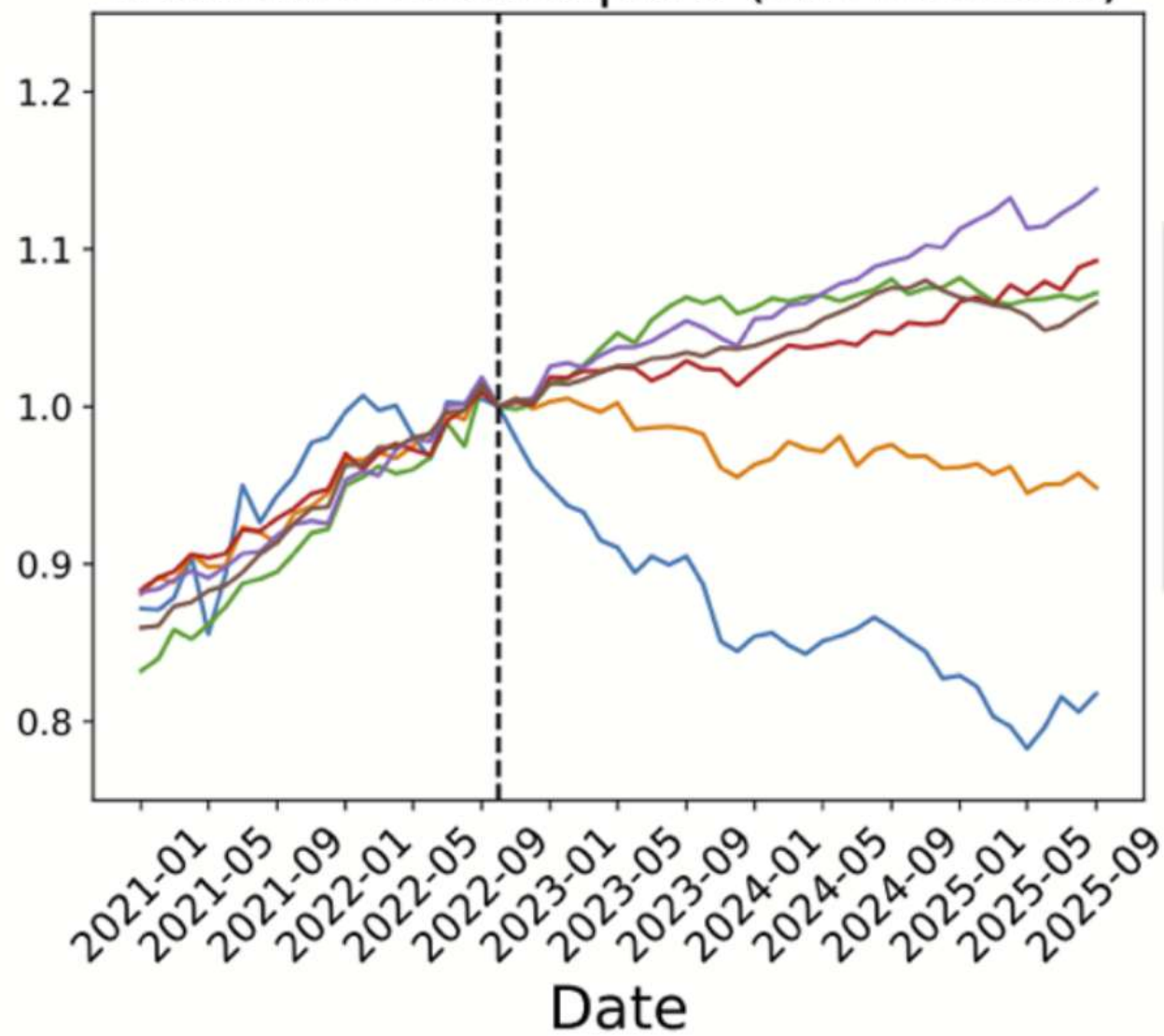
Humlum & Vestergaard, May 2025

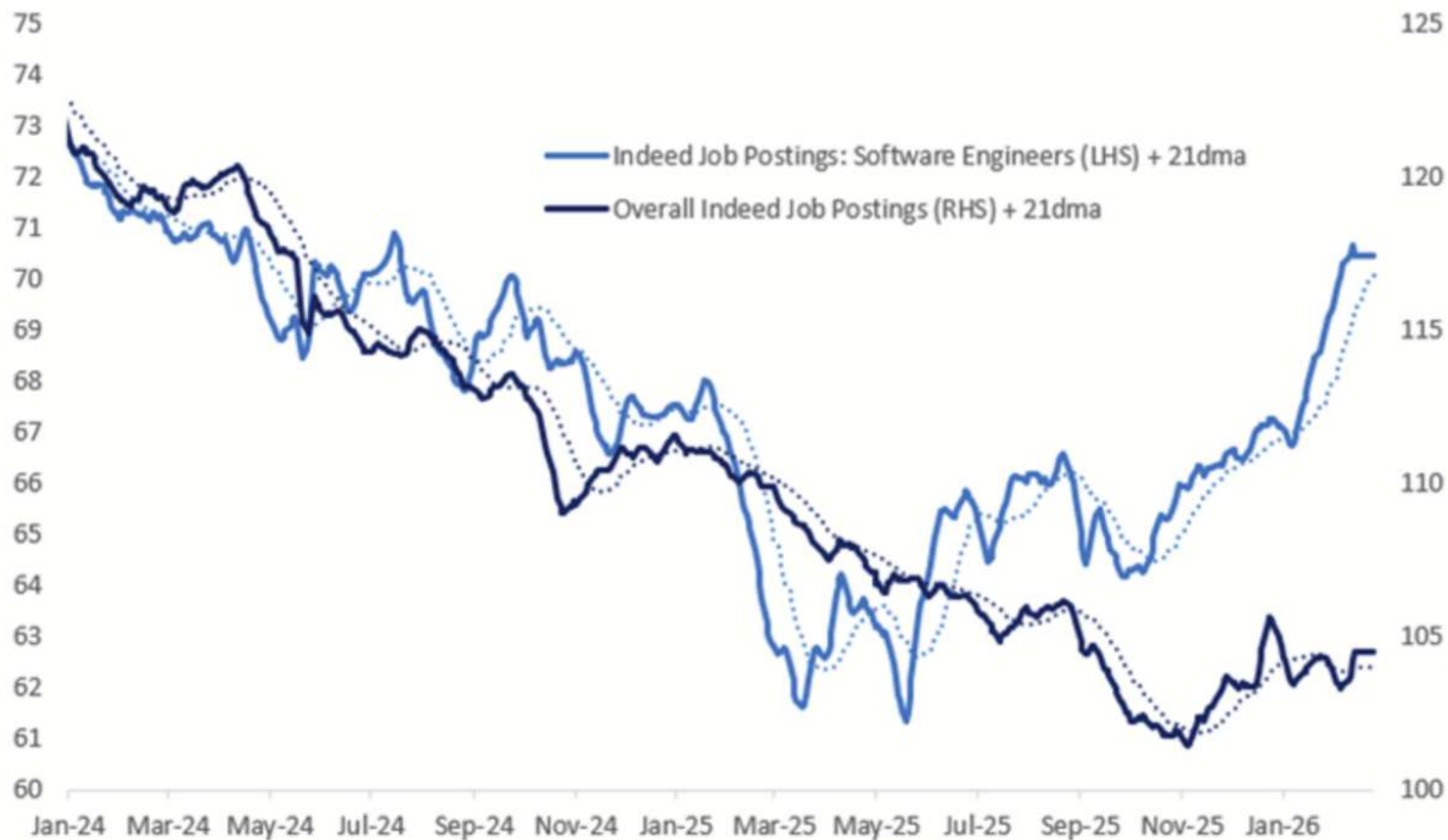
“Large Language Models, Small Labor Market Effects”

We examine the labor market effects of AI chatbots using two large-scale adoption surveys (late 2023 and 2024) covering 11 exposed occupations (25,000 workers, 7,000 workplaces), linked to matched employer-employee data in Denmark.

AI chatbots have had no significant impact on earnings or recorded hours in any occupation, with confidence intervals ruling out effects larger than 1%. Modest productivity gains (average time savings of 3%), combined with weak wage pass-through, help explain these limited labor market effects.

Headcount Over Time by Age Group
Software Developers (Normalized)

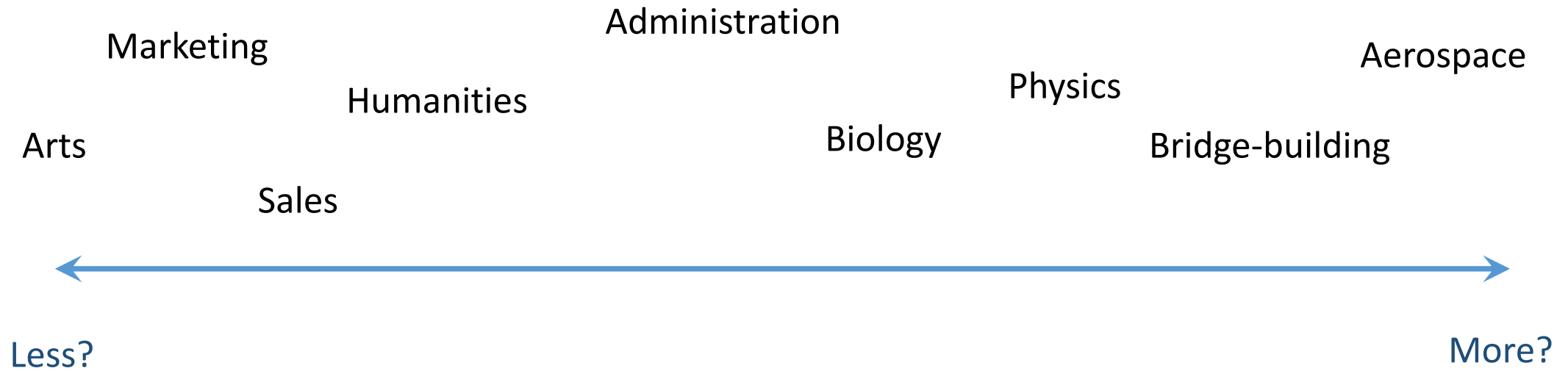




MIT study: The GenAI Divide: State of AI in Business 2025

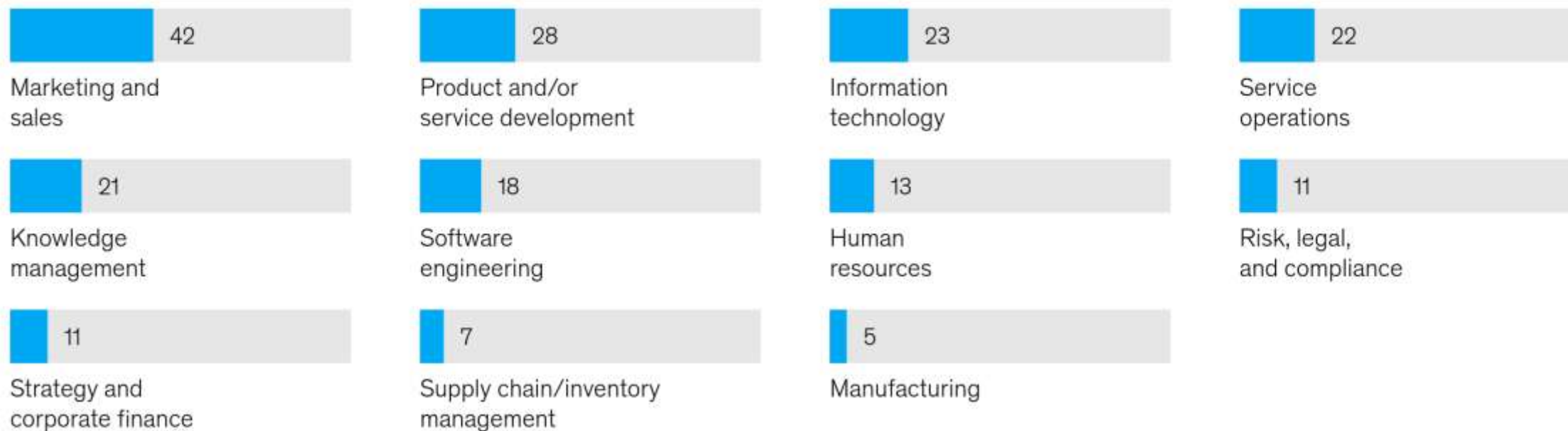
- Only 5% of organizations are seeing transformative returns. The vast majority—95%—report zero impact on profit and loss statements from formal AI investments.
- The core barrier to scaling is not infrastructure, regulation, or talent. It is learning. Most GenAI systems do not retain feedback, adapt to context, or improve over time.
- At the same time, employees in over 90% of companies regularly use personal AI tools for work.

Where does truth matter?



McKinsey report, March 2025

Use of gen AI by business function, % of respondents



Oletused

- A.I. areneb ennustamatutes suundades, ennustamatu kiirusega.
 - Pea kõik “tava-tööd” jäävad nähtavas tulevikus alles.
 - Vajalikest A.I. / IT oskustest on väga puudu.
 - A.I. võimestab tarkvara arendajaid.
 - Nõudlus tarkvara arenduse vastu kasvab.
 - Õpetamise struktuuri variatiivsust tasub suurendada.
- Näiteks: käivitame integreeritud erialasid lisaks hüperspetsialiseeritutele.