

The cost of AI

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Introduction

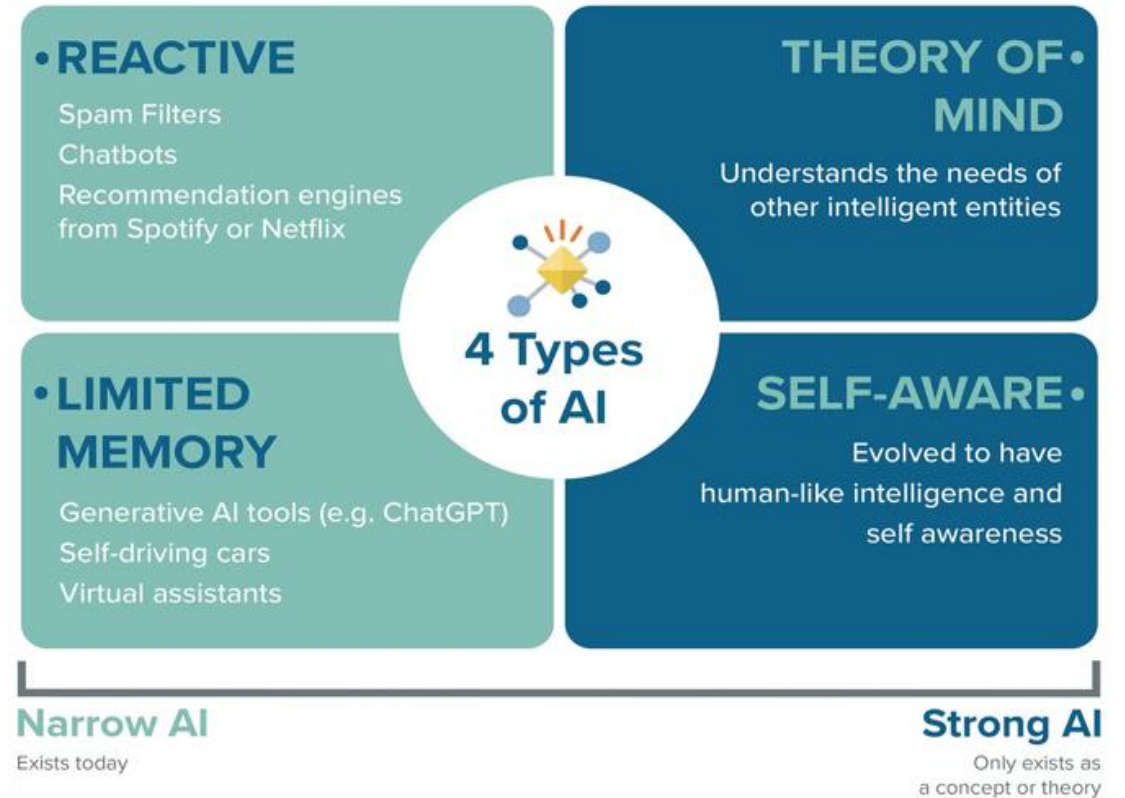
- The usual question about AI
 - How shall we use it?
- Often overlooked questions about AI
 - What is the cost of using it?
 - What is the benefit/ratio cost?
 - What are the implications behind its usage?
- Just like any tool
 - The potential to serve humanity exists, if used ***reasonably well***
 - It has a production cost in the equation
 - It also has consequences due to usage

The question we ***don't*** ask:

Under which conditions is it acceptable to use it?

AI definition

- Group of technologies that **process information** and **simulate** human activities
- Such activities include **understanding, learning, creating, reasoning, decision making, ...**
- Science aiming at automating the process of **learning** machines to **perform tasks**

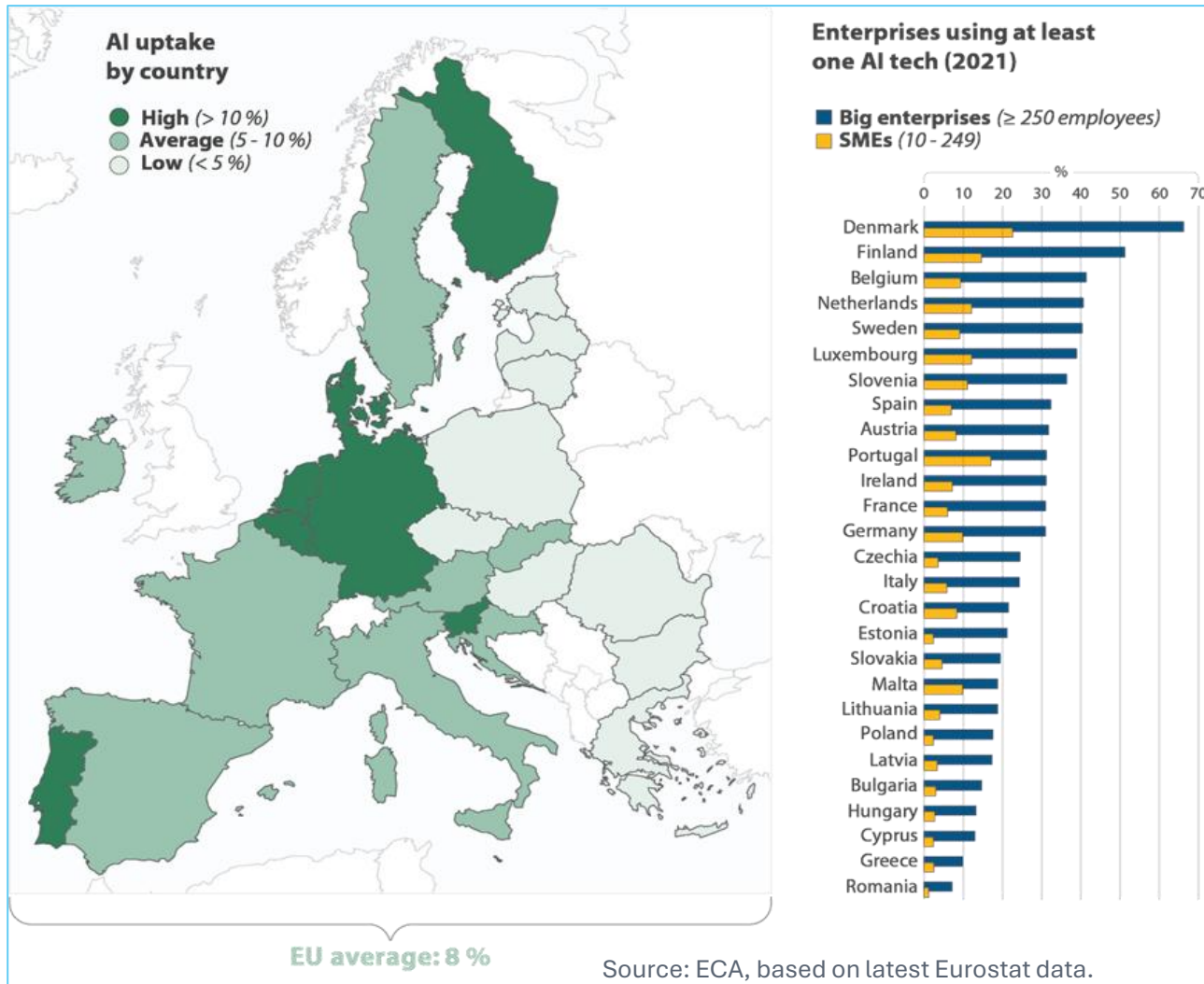


<https://www.aitechboss.com/types-of-ai/>

AI benefits

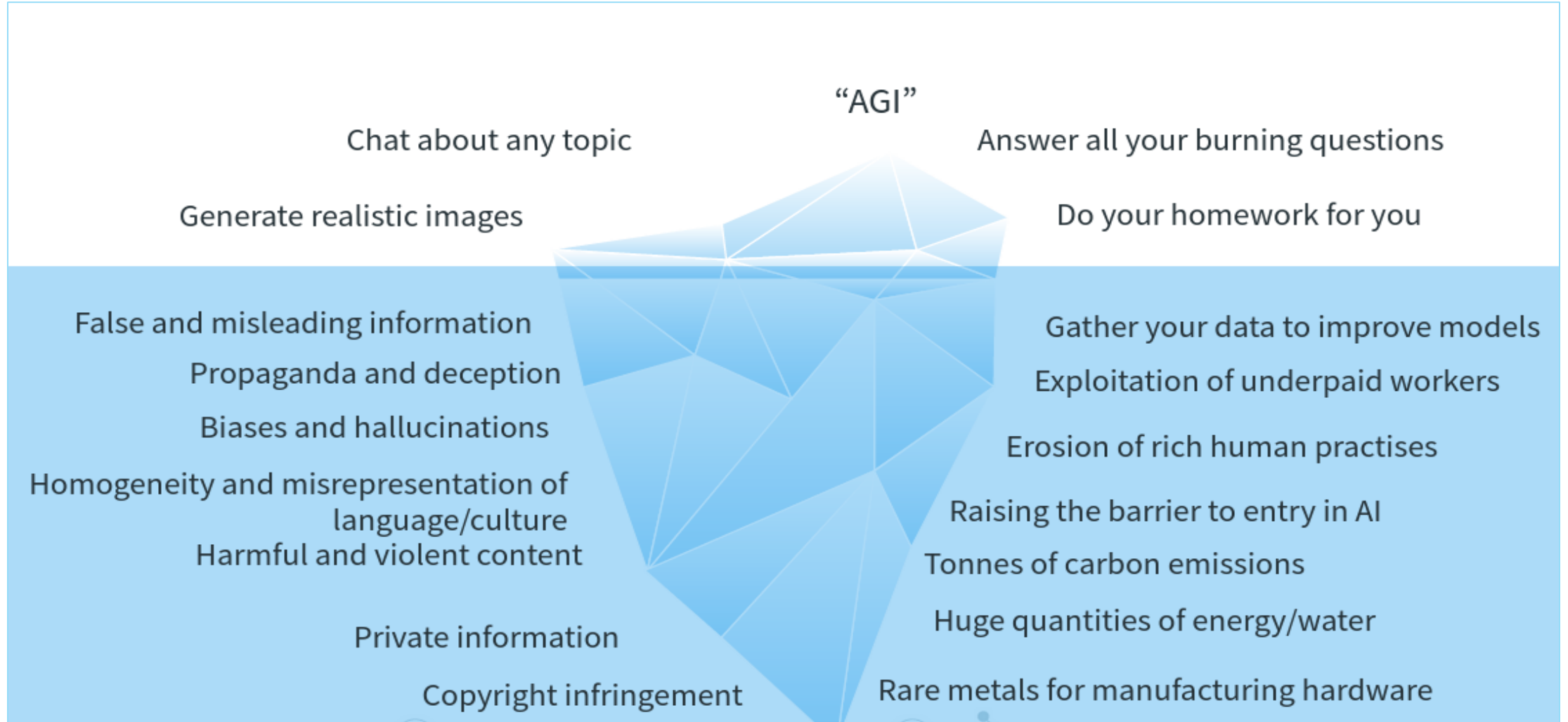
- Makes everyday tasks faster
 - Usually works like an assistant
 - Through interactive prompt
 - Simulates human dialogue
 - Provides some support
 - Needs to be double-checked
- Usual AI task examples
 - Summarize long documents
 - Write draft emails
 - Manage and schedule tasks
 - Help with decisions
 - Generate creative contents
 - Help with trip planning
 - Give short definitions

AI uptake in Europe



- **Massive adoption**
 - Large companies
 - Followed by SME
- **Not only companies**
 - 1/3 of the 16-74 population used AI in 2025
 - 25% for personal purposes
 - 15% of them for work
 - 9% for formal education

AI issues



AI cost: multiple dimensions

- Material resources
 - electricity, water, servers, chips, and data-centre infrastructure (de Vries, 2023; Li et al., 2025)
- Economical cost
 - low pay, economic dependency, outsourcing, surveillance, secrecy, and weak labour protections in parts of the AI data-work pipeline (Miceli & Posada, 2022; Miceli et al., 2024)
- Psychological damage
 - Data filtering work expose workers to significant mental-health risks (Miceli et al., 2024)
 - Also its usage can be damaging (Dang, Liu, 2025)

AI cost: some numbers

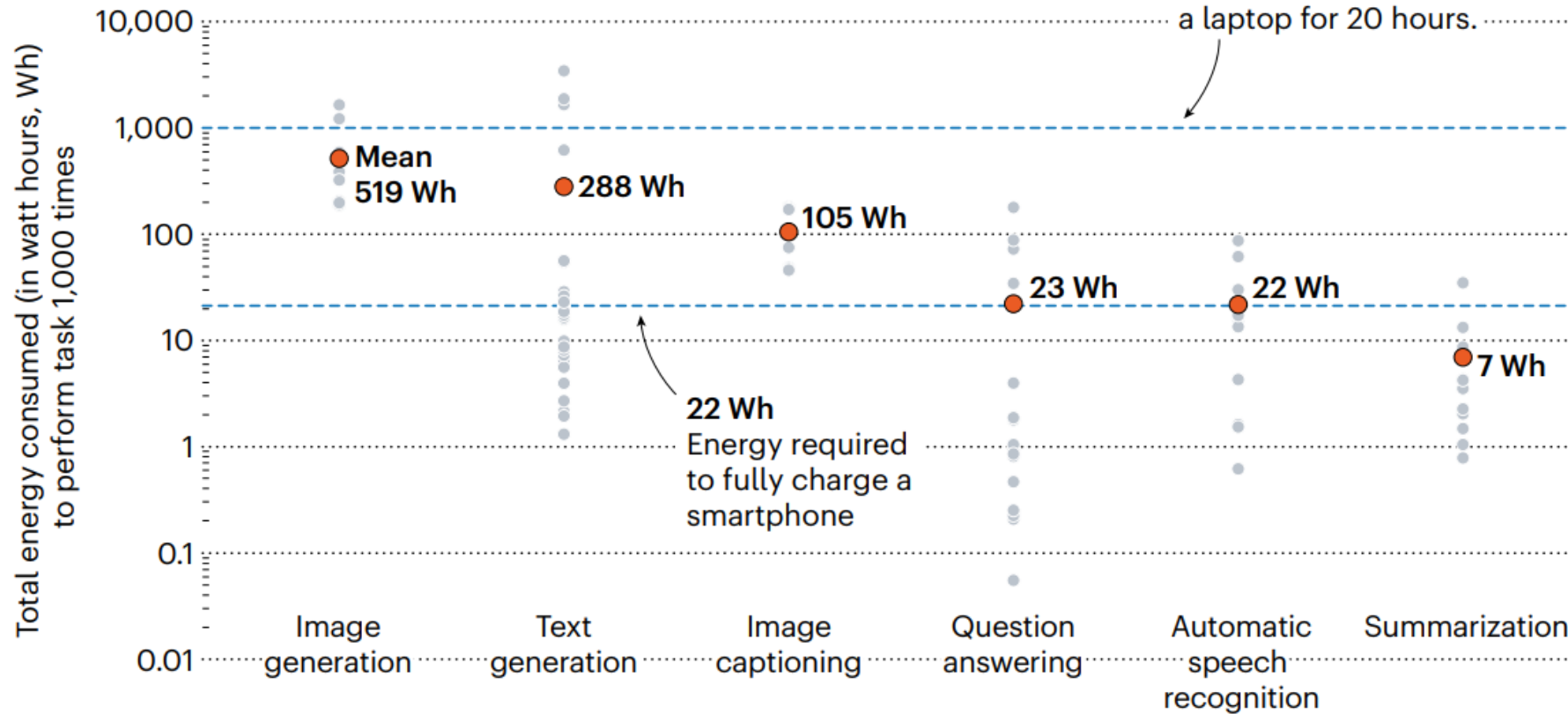
- Carbon footprint and water footprint
 - 1 short GPT-4o query consumes 0.43 Wh
 - Scale this to 700 million queries/day, you get...
- Huge environmental impact
 - electricity use comparable to 35,000 U.S. homes
 - freshwater evaporation matching the annual drinking needs of 1.2 million people
 - carbon emissions requiring a Chicago-sized forest to offset
- Estimation for GPT-3
 - Around 30 prompts consumes around ½ liter of water
- Energy needed for a prompt vs email
 - Estimated 10 to 100 times higher for a prompt

Energy cost comparison

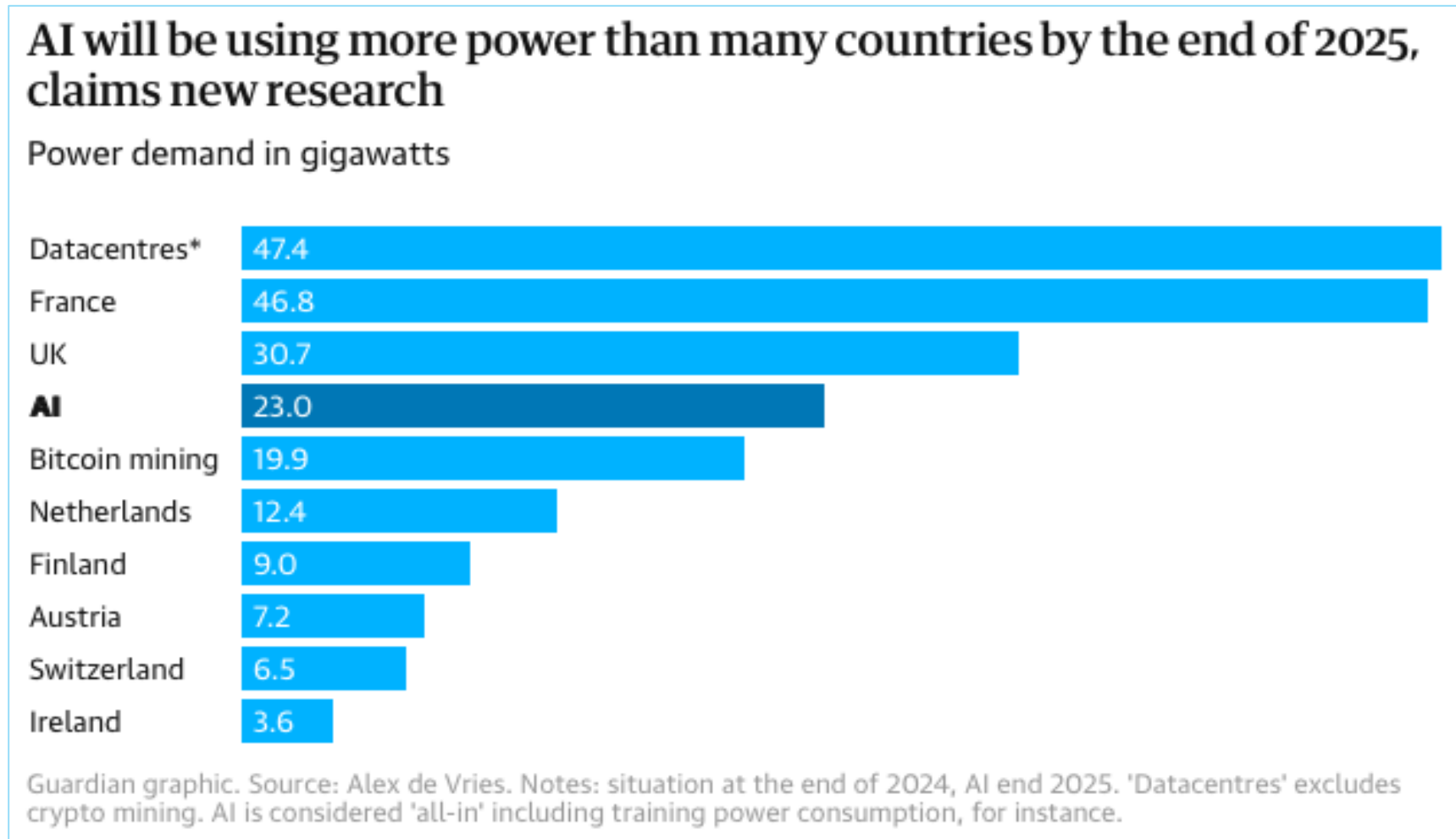
HOW MUCH ENERGY DOES AI USE?

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The AI Energy Score project tested dozens of artificial-intelligence models to estimate how much energy they consume when performing various tasks. Plotting the energy required to perform a task 1,000 times shows that energy use varies greatly depending on the task and the model.



Cost comparison with countries



Data center electricity usage and AI

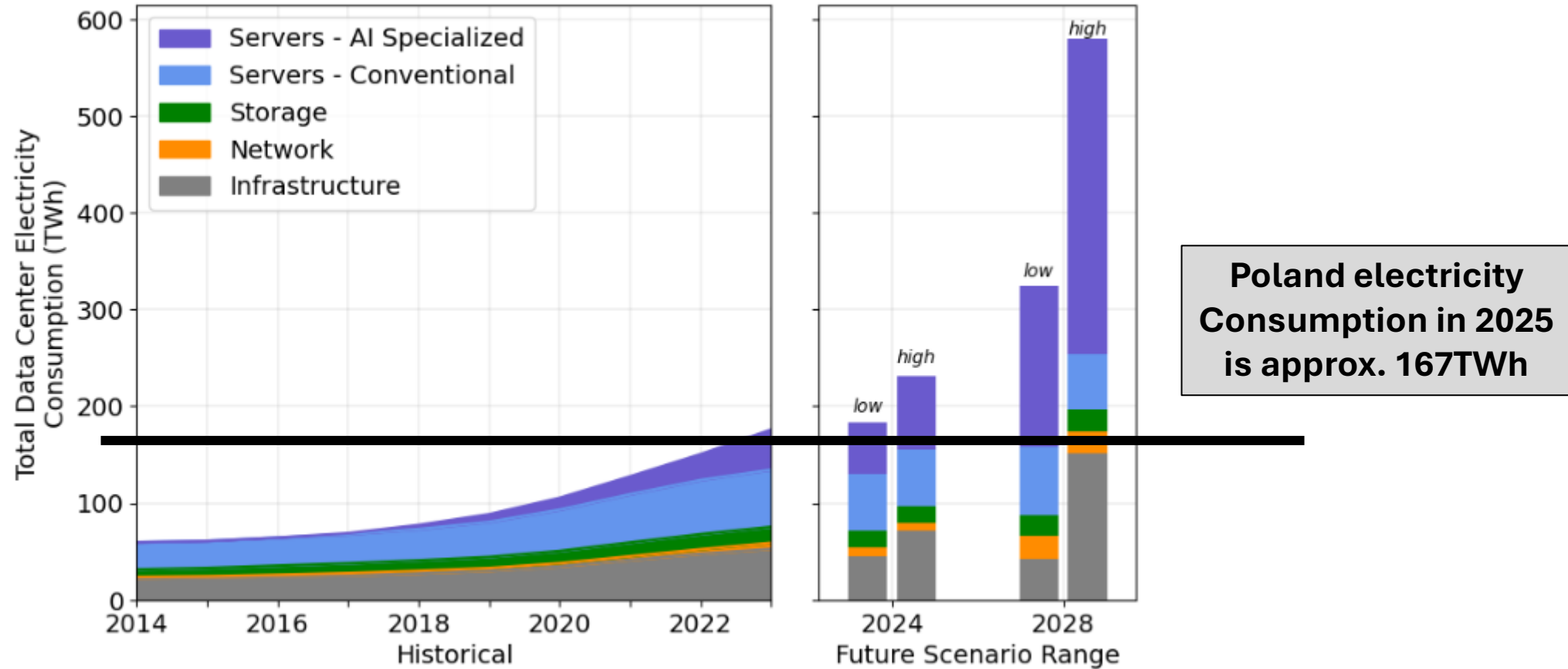
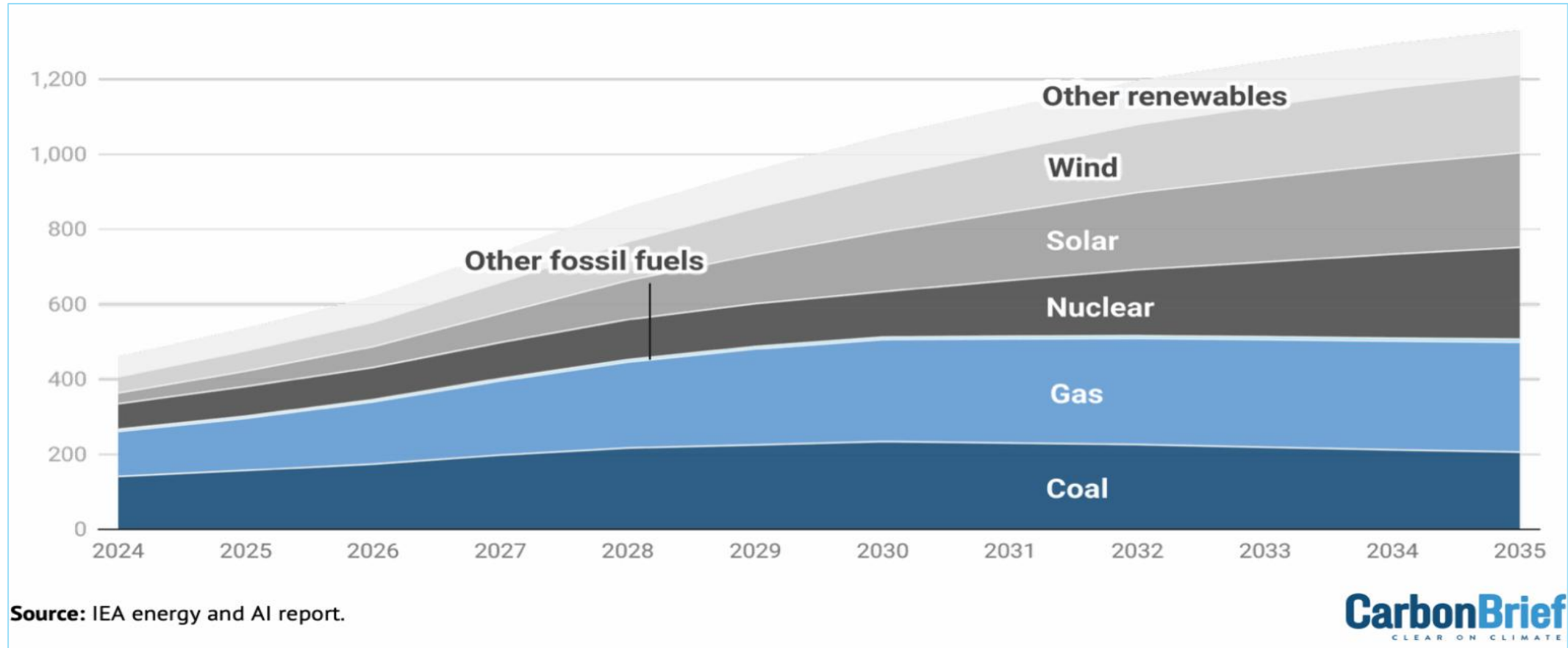


Figure 5.6. Total data center electricity use from 2014 through 2028 by equipment type.

Data centers and fossil fuels



Carbon footprint comparisons

in lbs of CO2 equivalent

Roundtrip flight b/w NY and SF (1 passenger)

1,984

Human life (avg. 1 year)

11,023

American life (avg. 1 year)

36,156

US car including fuel (avg. 1 lifetime)

126,000

Transformer (213M parameters) w/
neural architecture search

626,155

Chart: MIT Technology Review • Source: Strubell et al. • [Created with Datawrapper](#)

1 lb \approx 0.45 kg \rightarrow 1,984 lbs \approx 900 kg

Processing the AI data

- Human tasks in AI
 - Before training: data generation, collection, filtering, annotation
 - During training: assessing model responses (true/false)
 - After training: Impersonation: replacing AI in some contexts
 - Yes, they do ask humans to pretend and reply like a machine.
- Those tasks remains human-based
 - They are outsourced in platforms, using underpaid workers
 - With NDAs, surveillance, isolation, less support (healthcare, pension)
 - Vulnerable populations are targeted: refugees, prisoners, children
 - (Miceli, 2024)

AI usage: the psychological cost

- The AI bias
 - Conveys the a certain bias that the data it builds on has
 - Online data producers do **not** reflect the world population
- Dimensions of the bias
 - Imperialist worldview, white supremacist views, gender-biased views, (you-name-it)-phobic views...
 - Are more present in the generated knowledge model (Katz, 2020)
- Many examples in the news
 - Such as the Dutch authority scandal
 - An AI-assisted program wrongly accused 26000 families of fraud for child support declaration, asked for unfair reimbursements

AI usage: the psychological cost

- Psychological dependency and attachment formation
 - Users often anthropomorphize AI systems
- Parasocial attachment
 - Leads to delusional thinking, emotional dysregulation, and social withdrawal
- Crisis incidents and harmful outcomes
 - Cognitive impairment and addictive behaviors linked to prolonged AI use
- Heightened vulnerability among specific populations
 - Adolescents, elderly adults, and individuals with mental illness
- Need for validated diagnostic criteria, clinician training, ethical oversight, and regulatory protection

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Conclusion

- The decision to use AI belongs to the user
 - The costs and consequence of AI must be known
- The objective of this set of slides
 - Is to inform about what is often silenced
 - To contribute to better clarity about this tool
 - To raise awareness about its impact
- Feel free to improve these slides
 - Just send me back your improved version, I might update this one

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