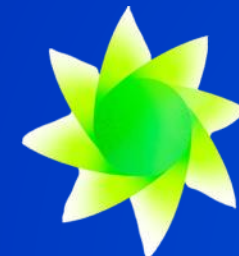




AI and environment: an impossible equation ?

GREGORY LEBOURG

27.11.25



Coriolis Seminars
for the Environment

Agenda

01

What is Cloud AI ?

Materiality of the infrastructure

02

Assessing the AI impact

Where are we ?

03

Data centres utilities demand

Water & electricity projections, what infrastructure technological solutions ?

04

Beyond utilities

Manufacturing emissions, abiotic resources and land usage, what mitigation levers ?

05

Users and developers role

For a responsible ecosystem

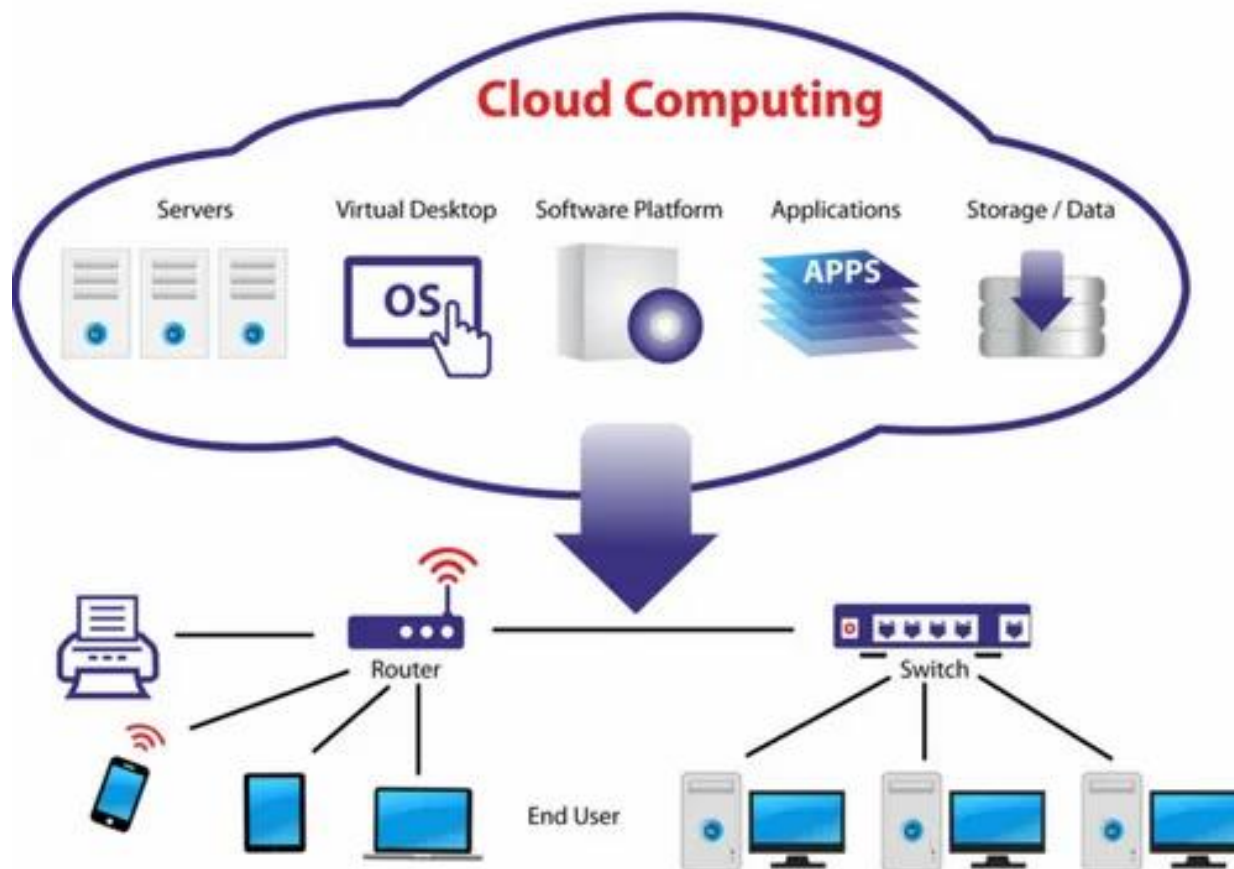
01

What is Cloud AI ?



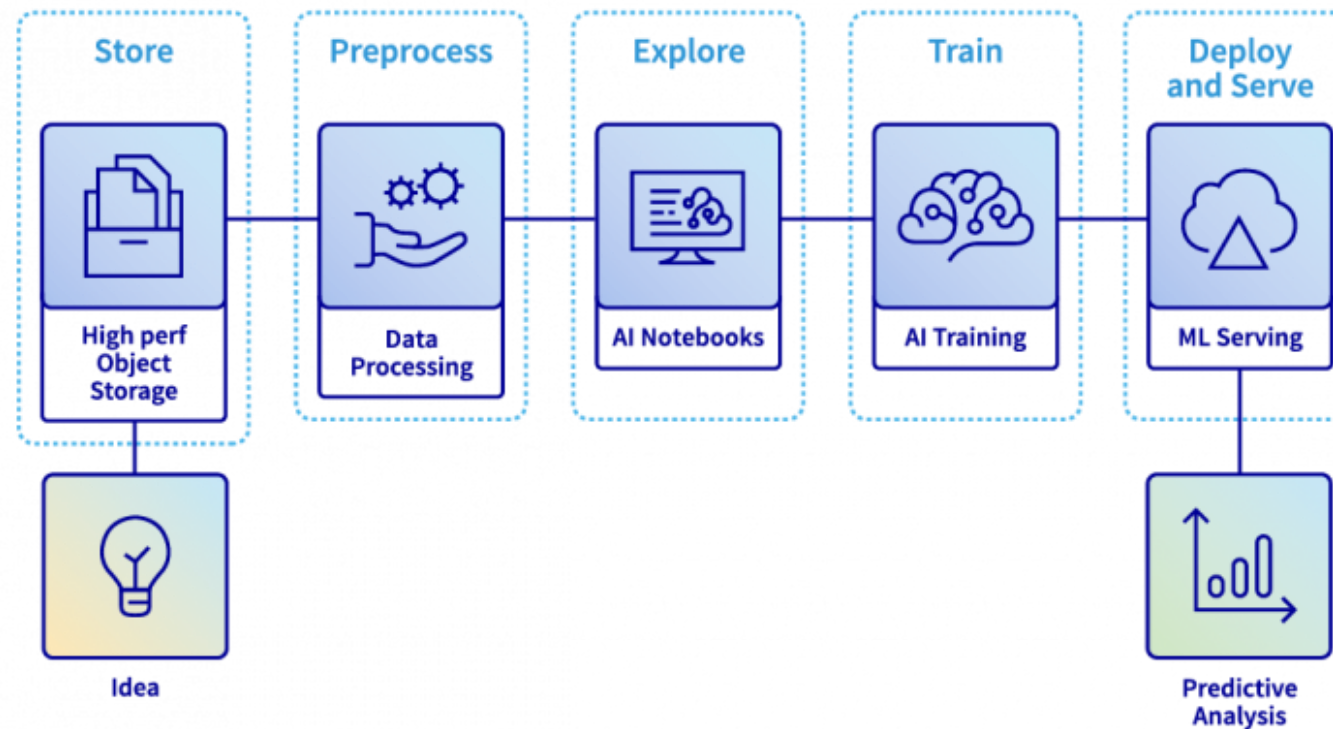
What is Cloud ?

Cloud refers to a system that provides access to IT services (storage, computing, software) through the internet or private networks



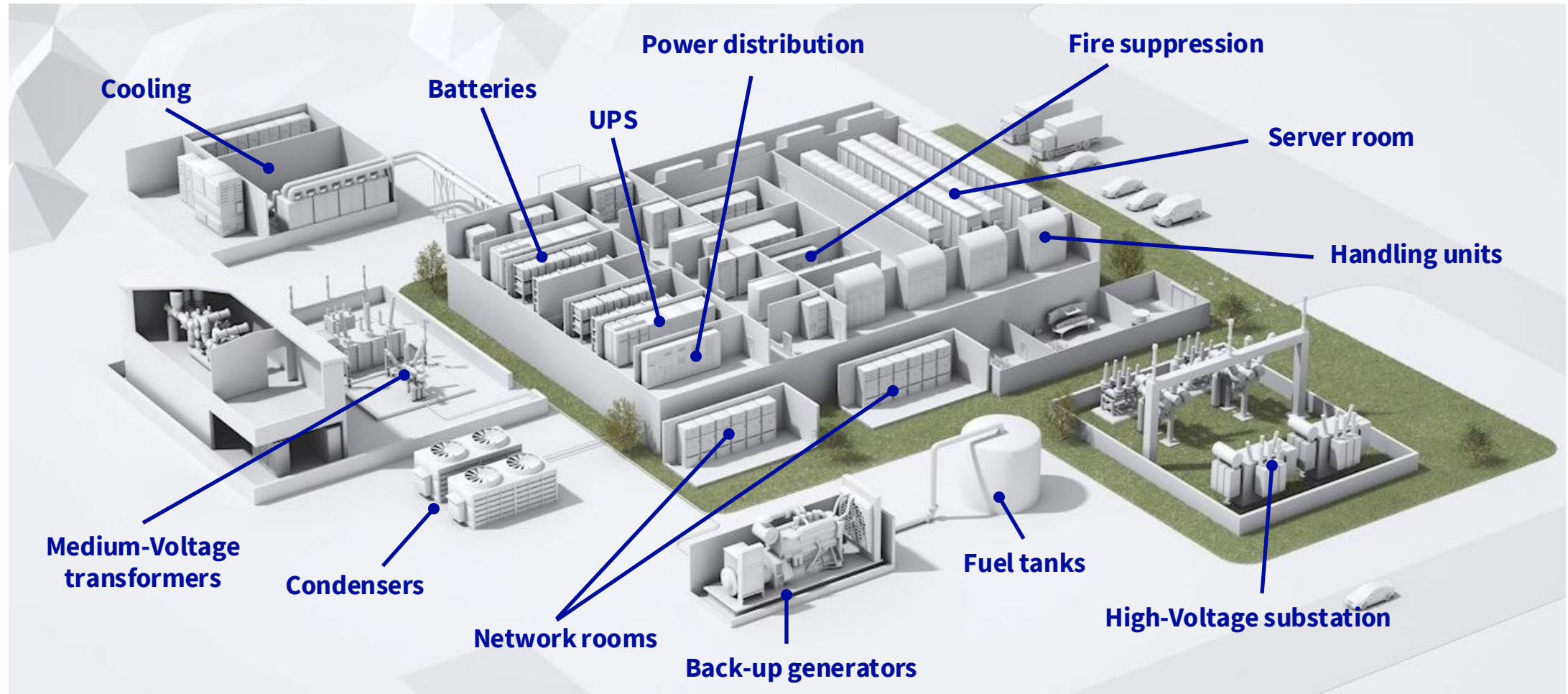
What is Cloud AI ?

Cloud AI refers to the integration of artificial intelligence in a Public Cloud platform. It enables organisations to leverage enormous computing power and advanced AI processes without depending on costly, inefficient on-premises servers.



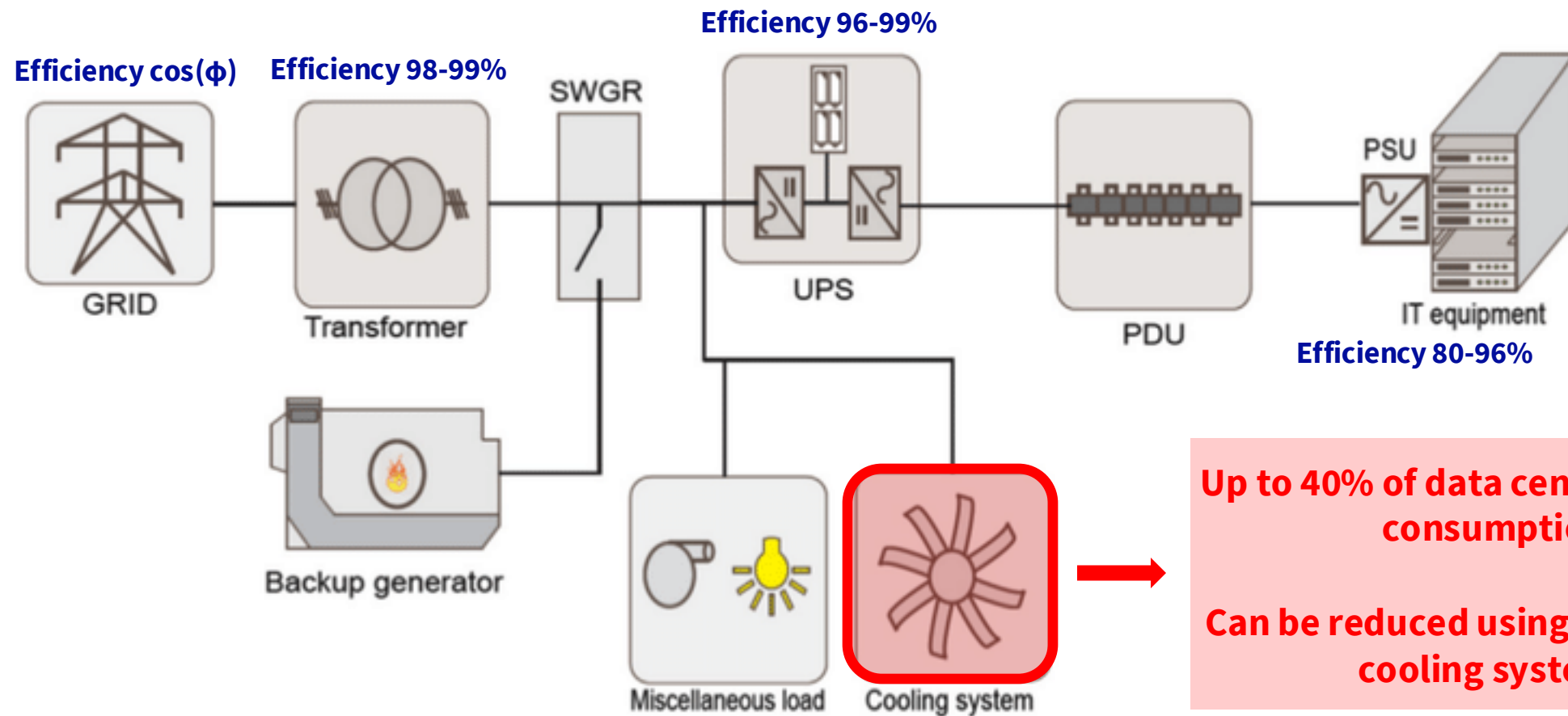
The Cloud platforms are hosted in gigantic technical infrastructures

Data centres = land + buildings + industrial gears + utilities (energy/water/telco)



Whatever the design, it's all about chasing the inefficiency

2 KPIs to be optimized **PUE** (Power Usage Effectiveness) / **WUE** (Water Usage Effectiveness)

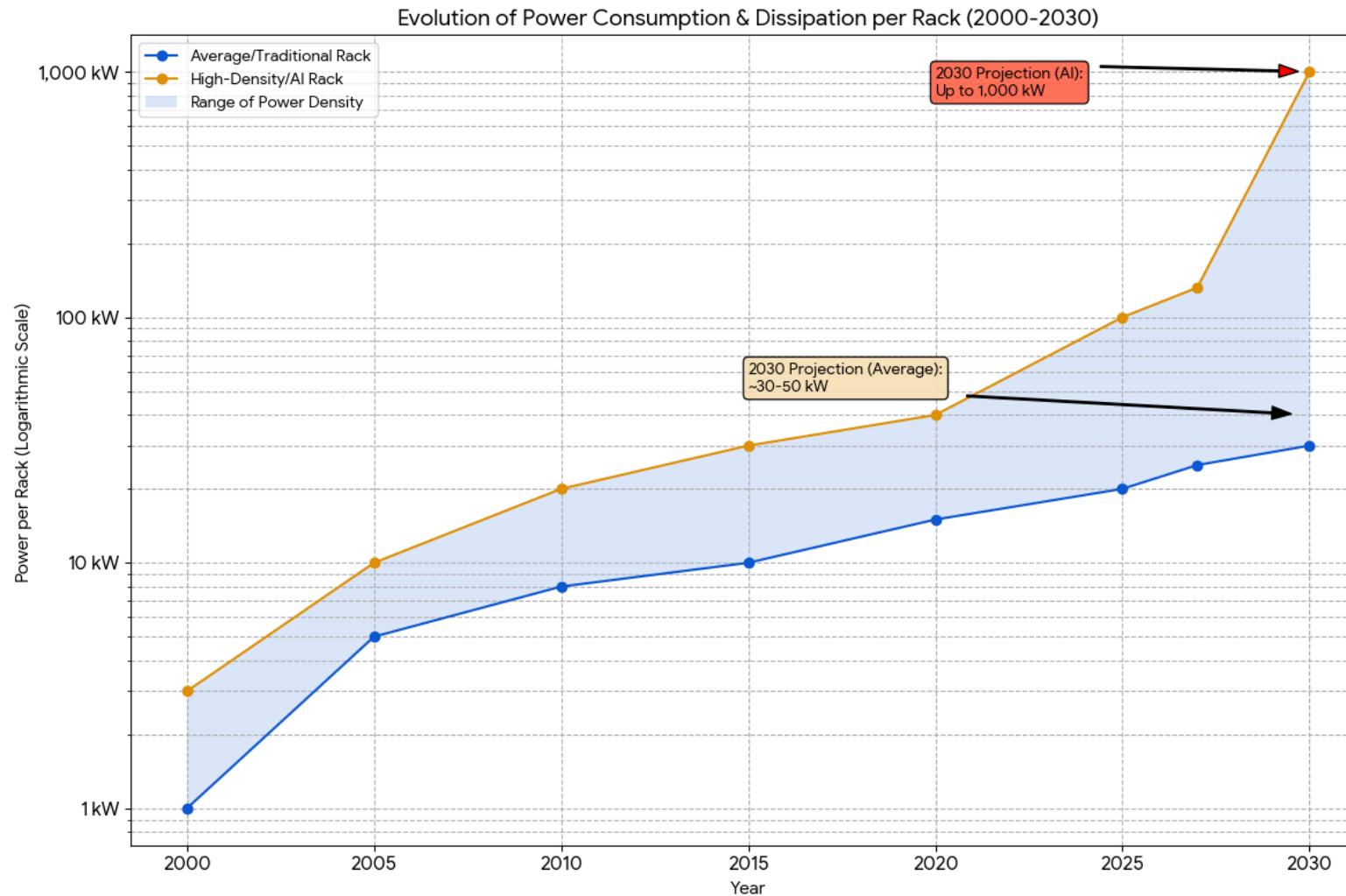


Up to 40% of data centre electrical consumption

Can be reduced using evaporative cooling systems

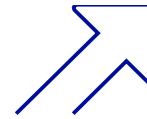
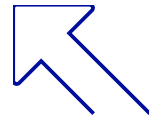
The Cloud AI is built on GPU based hardware platforms

Power density reaches a level never seen before

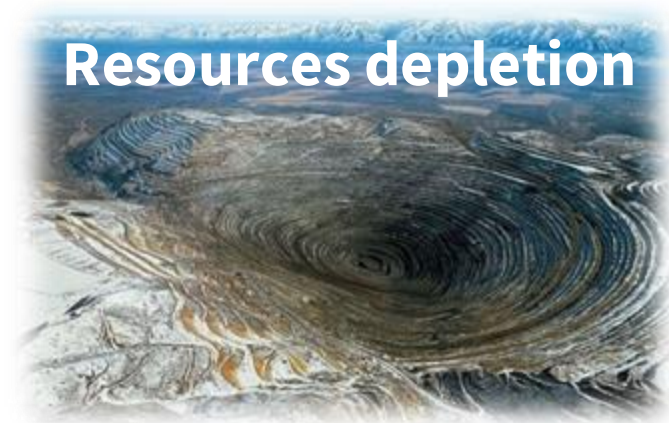
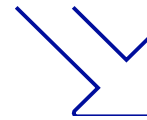
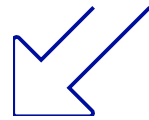


Source – Lennox Data Center Solutions

Consequence : Cloud AI is adding to 4 main environmental issues



AI



02

Assessing the AI impact



Only one can solve a problem by assessing it correctly...

More and more research papers being published



Our contribution to a global environmental standard for AI

Company

Jul 22, 2025 | Mistral AI



Google Cloud

Measuring the environmental impact of AI inference

Infrastructure

How much energy does Google's AI use? We did the math

August 21, 2025

... but we are not there yet !

Misaligned methodologies and wilful bad faith blur the picture



1.14 gCO2e /prompt



45 ml /prompt



0.03 gCO2e /prompt

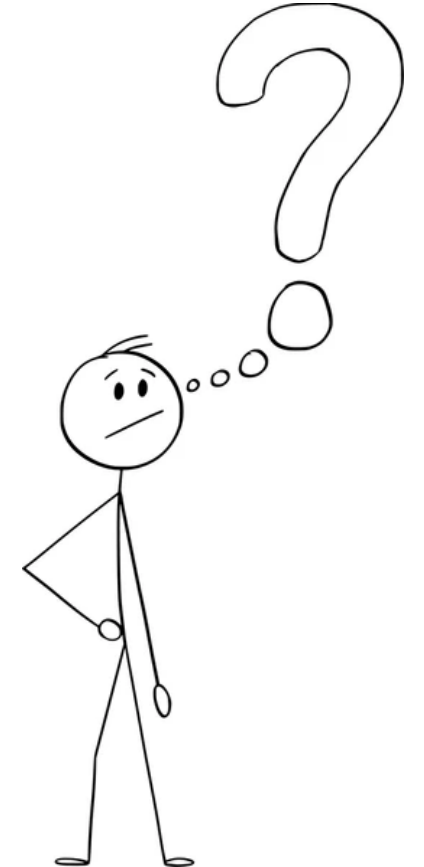


0.26 ml /prompt



÷ 38

÷ 173












Conclusion : Never trust the numbers before checking what's behind

Some in-depth benchmarks are available

2. Benchmark : Carbon calculators from main Cloud providers

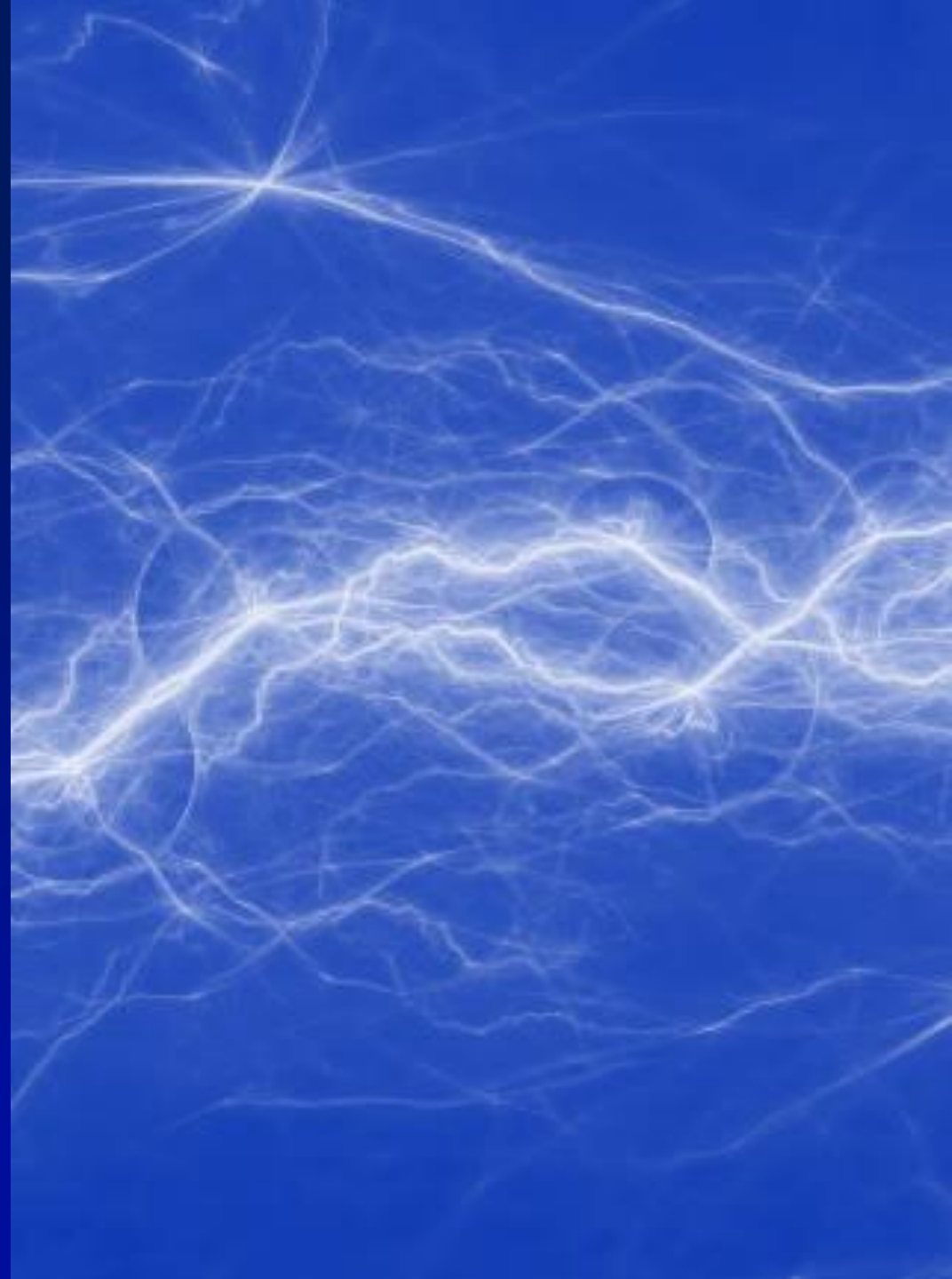


| | | Scope 1 | Scope 2 | Scope 3 | | | |
|--|---|---------------|--|---------------------------------|---------------------------------|--------------|------------------------|
| | | Fossil energy | Electricity | Servers & equipment | Data Center | Internet | Cloud provider's Teams |
| ICT Sector Guidance |  | To include | To include | Depreciation based on life span | Optional | To include | To include |
| PCR Datacenter et Cloud |  | To include | Location Based | Depreciation based on life span | Depreciation based on life span | To exclude | To include Partially |
| | | | | | | | |
|  | | Included | Market & Location based | 1/5 from commissioning | Included | included | Included |
| | | | | | | | |
|  | | Included | Location based | 1/6 from commissioning | Included | included | Included |
|  Outil Open Source CloudAssess | | Included | Electrical mix and location to specify | Lifespan to specify | Included | not included | Partially included |
|  Google Cloud Platform | | Included | Market & Location based | 1/4 from commissioning | Included | not included | Partially included |
|  | | Included | Market based, Renewable to 0 | 1/6 from commissioning | not included | included | not included |
|   | | Included | Market based Renewable to 0 | not included | not included | not included | not included |

Source – IJO study July 2025

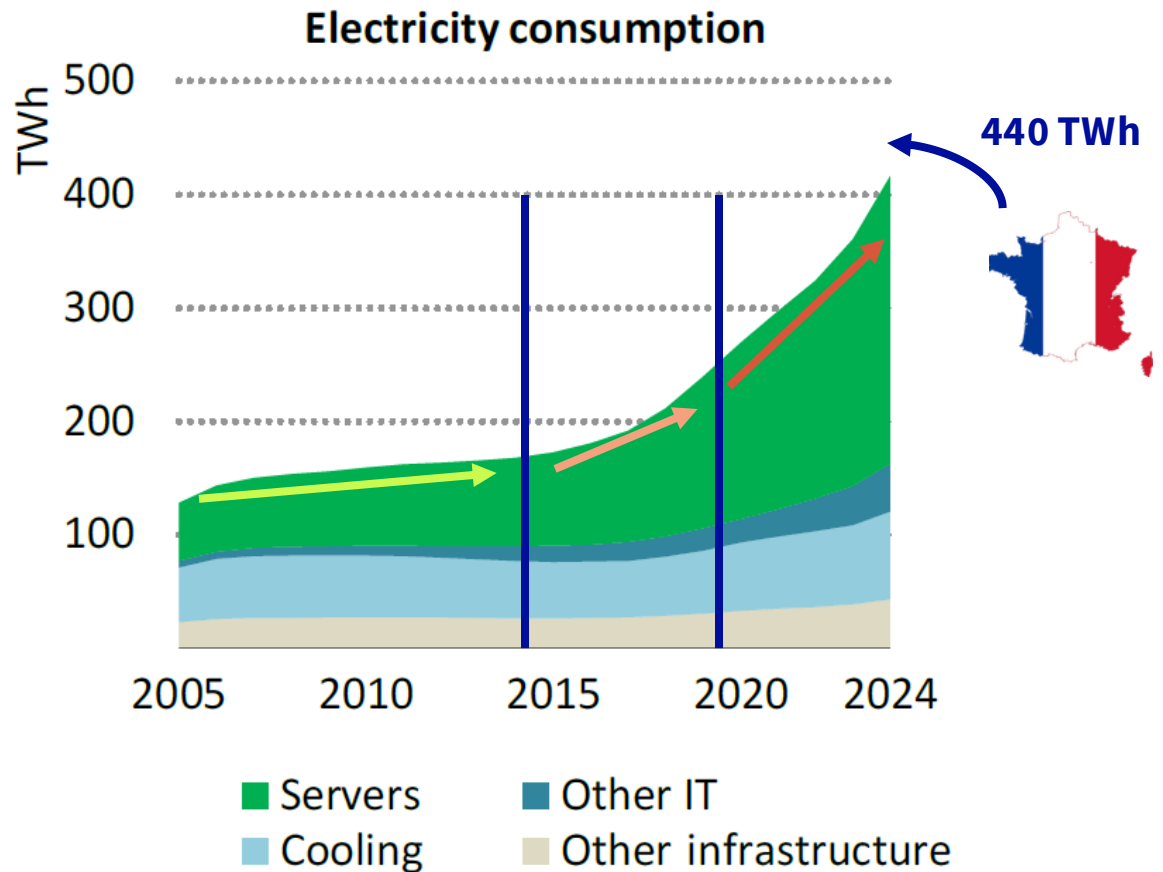
03

Data centres
utilities demand



Electricity - Let's have a look in the driving mirror

From 165 TWh in 2014, data centres energy demand has increased up to 420 TWh in 2024 (cryptocurrencies excluded)

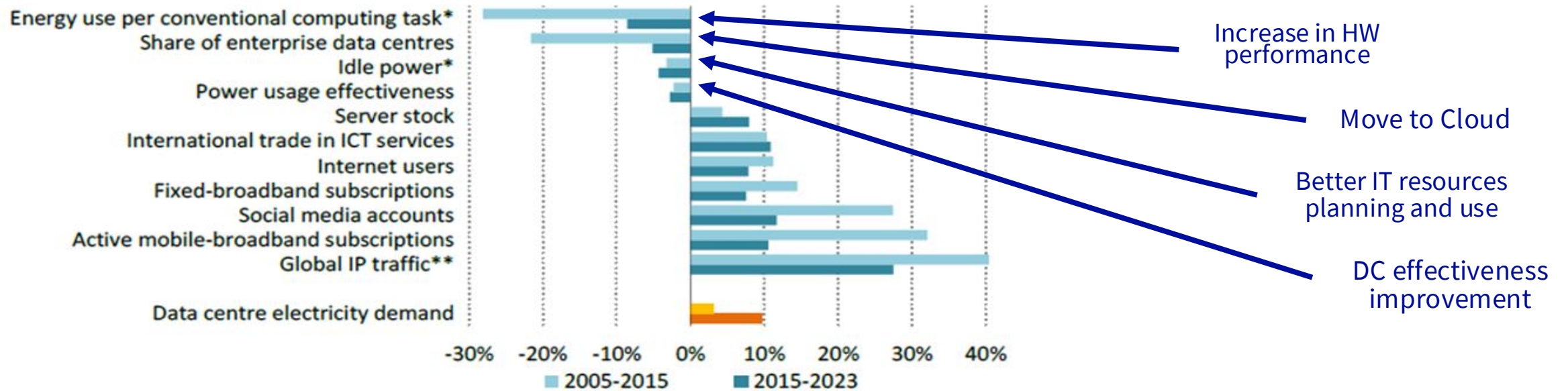


Source – « Energy and AI », IEA, 2025

IEA. CC BY 4.0.

Explanation : Increase in usage has overpassed the efficiency gain

Even more so that levers effects tend to decrease over time



IEA. CC BY 4.0.

Robust service demand growth, an acceleration in the total number of servers and a slowdown in some efficiency indicators led to faster electricity consumption growth

* Data starts in 2007. ** Data ends in 2022, estimated for 2022.

Source – « Energy and AI », IEA, 2025

Trend : AI demand should be mitigated by chips efficiency improvement...

GPUs keep on following Koomey's law

| GPU | TDP (W) ⁷ | TFLOPS ⁸ (Training) | Performance over V100 | TOPS ⁹ (Inference) | Performance over V100 |
|----------------|----------------------|-----------------------------------|--------------------------|----------------------------------|--------------------------|
| V100 SXM2 32GB | 300 | 15.7 | 1X | 62 | 1X |
| A100 SXM 80GB | 400 | 156 | 10X | 624 | 10X |
| H100 SXM 80GB | 700 | 500 | 32X | 2,000 | 32X |
| B200 SXM 180GB | 1,000 | 1,125 | 72X | 4,500 | 73X |
| B300 SXM 288GB | 1,400 | 1,880 | 120X | 7,500 | 121X |

Source – Schneider Electric SPD_WP110_EN V3

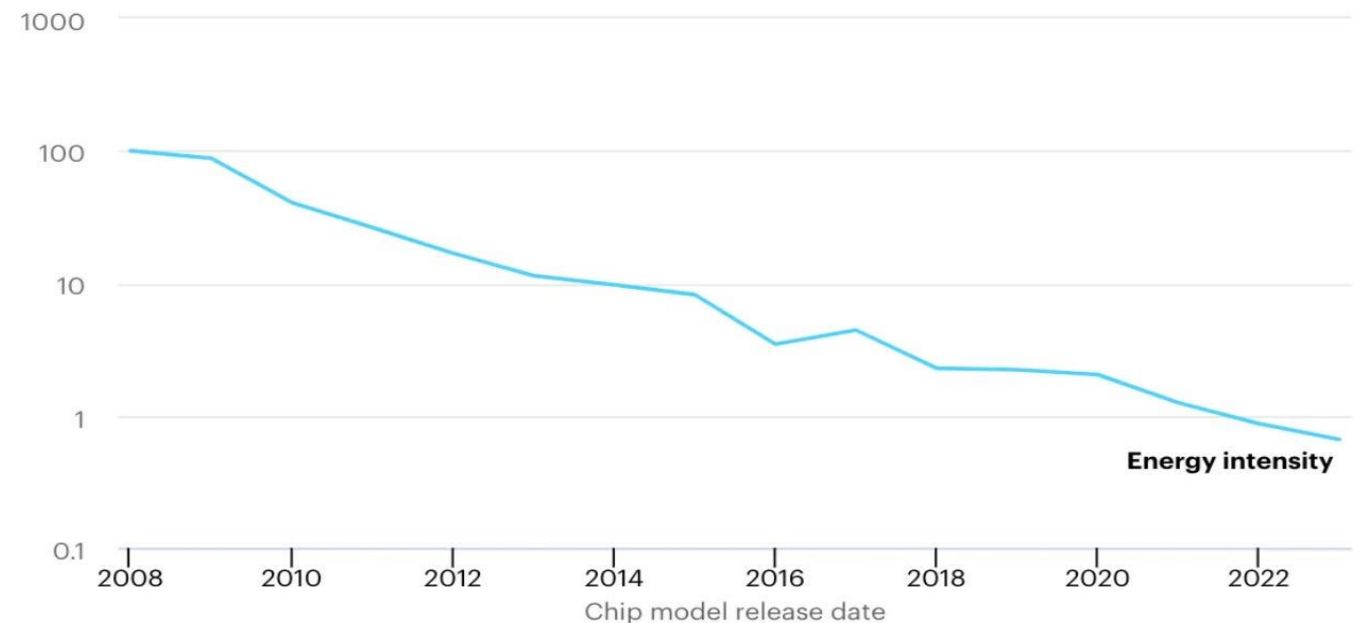
TDP (Thermal Design Power)

TFLOPS (Tera floating-point operations per second)

TOPS (Trillion of operations per second)

Efficiency improvement of AI related computer chips, 2008-2023

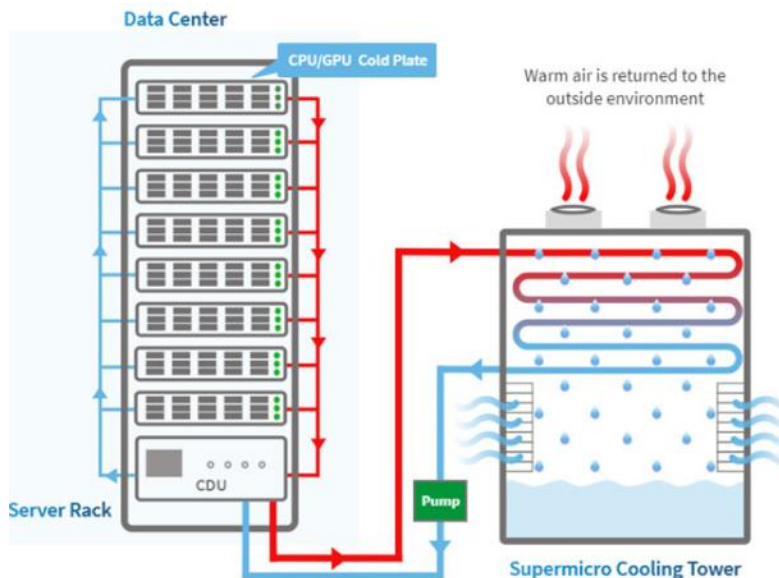
Index of energy intensity of AI computer chips (2008=100, log scale)



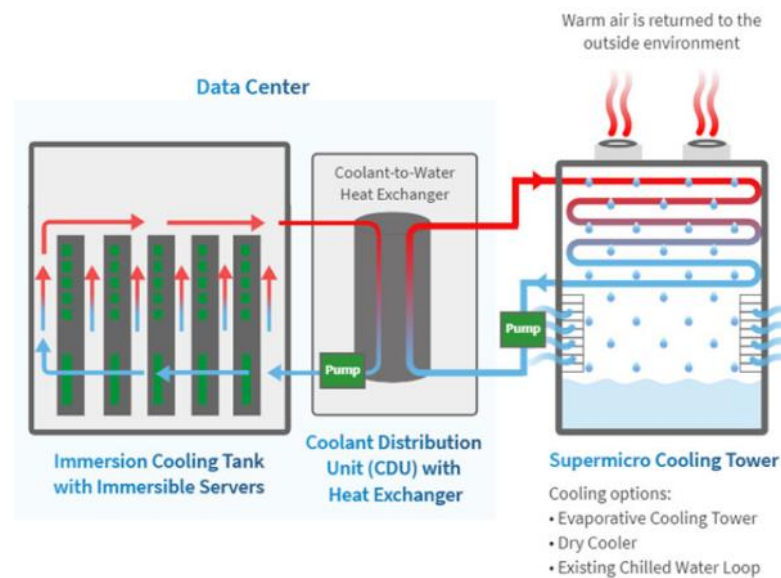
IEA. Licence: CC BY 4.0

... together with the generalization of liquid cooling for AI platforms

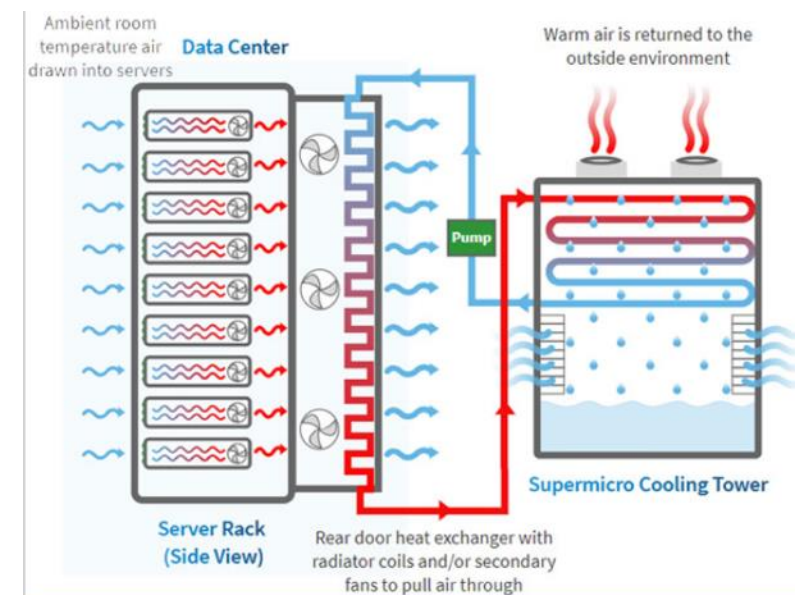
Massive adoption of liquid cooling for GPU will dramatically improve the PUE (Power Usage Effectiveness)



DIRECT TO CHIP



IMMERSIVE COOLING



REAR DOOR HEAT EXCHANGER

Illustration : OVHcloud solution and its benefit on the PUE

Our proprietary technology allows to keep up with the rack power increase while ensuring a high-power effectiveness

**500 000 water cooled servers
46 Data centres**

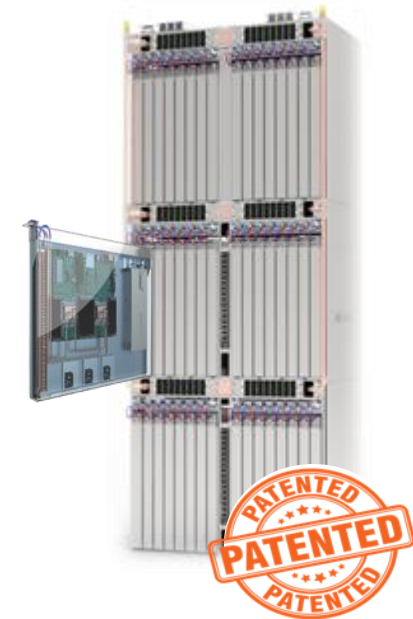
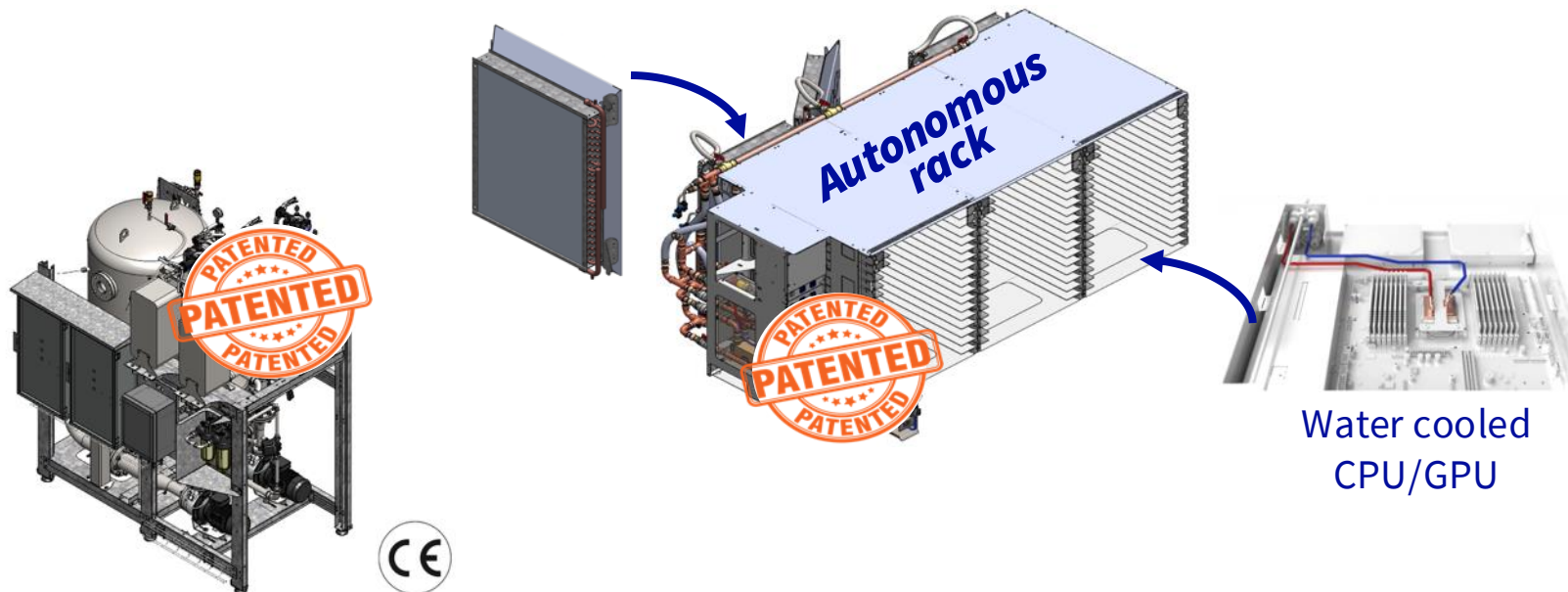
Introduced at scale 23 years ago
(100+ patents)

Worldwide Power Effectiveness

PUE* = 1.24 (1.26 previous year)
(vs 1.56 industry average)

Latest Designs Performance

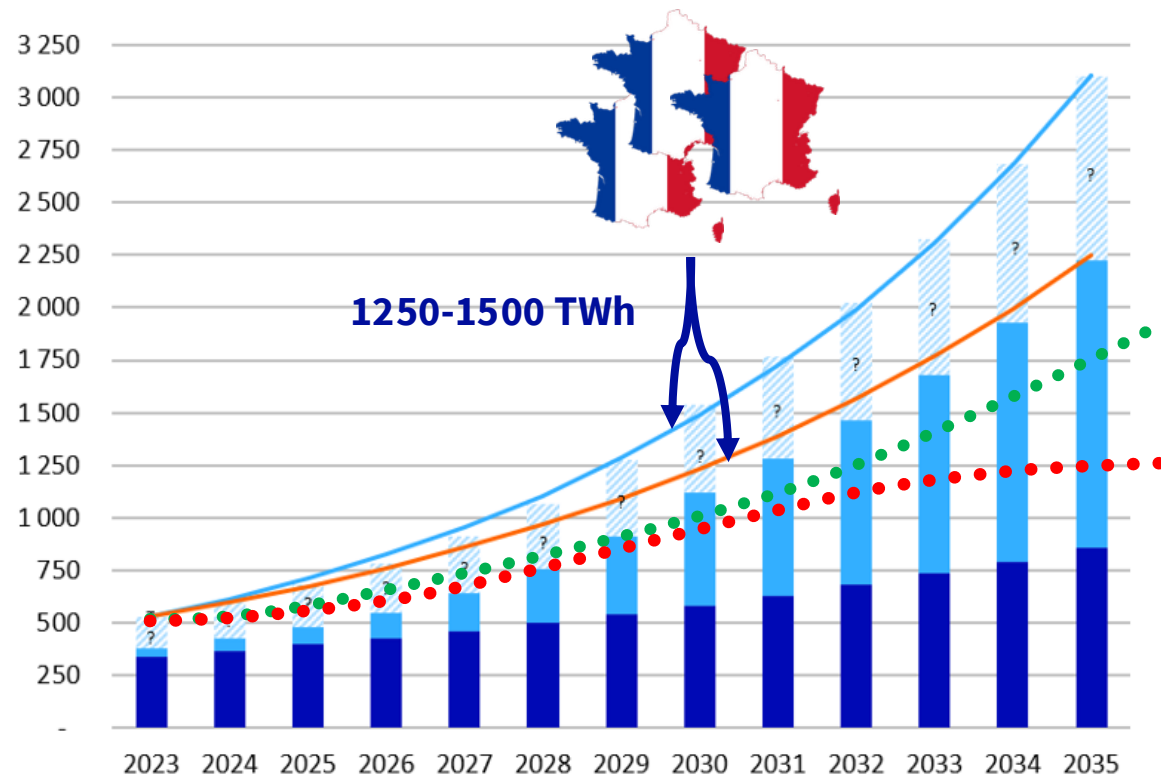
PUE as low as 1.10



** ISO 30134 audited*


Electricity : Let's have a look ahead but with critical thinking

Latest published forecasts are showing that energy demand is still accelerating pulling up the usage phase GHG emissions



Source – The Shift Project 2025


Three dynamics


 Cryptocurrencies

 Generative AI


 Traditional usages




Two scenarios

 Massive AI adoption and use of all available compute capacity

 Trend-based scenario
YoY 13 % (2019-2024 value)

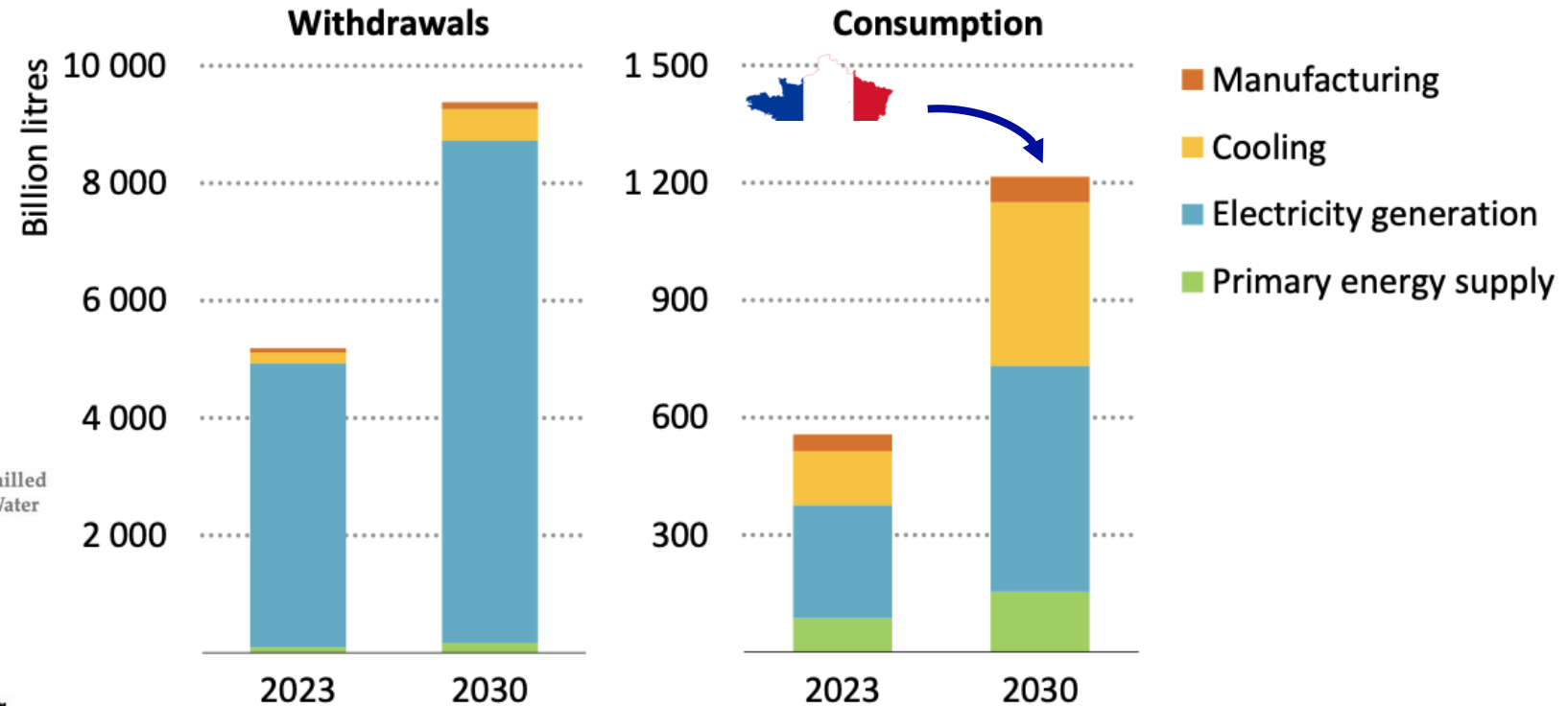
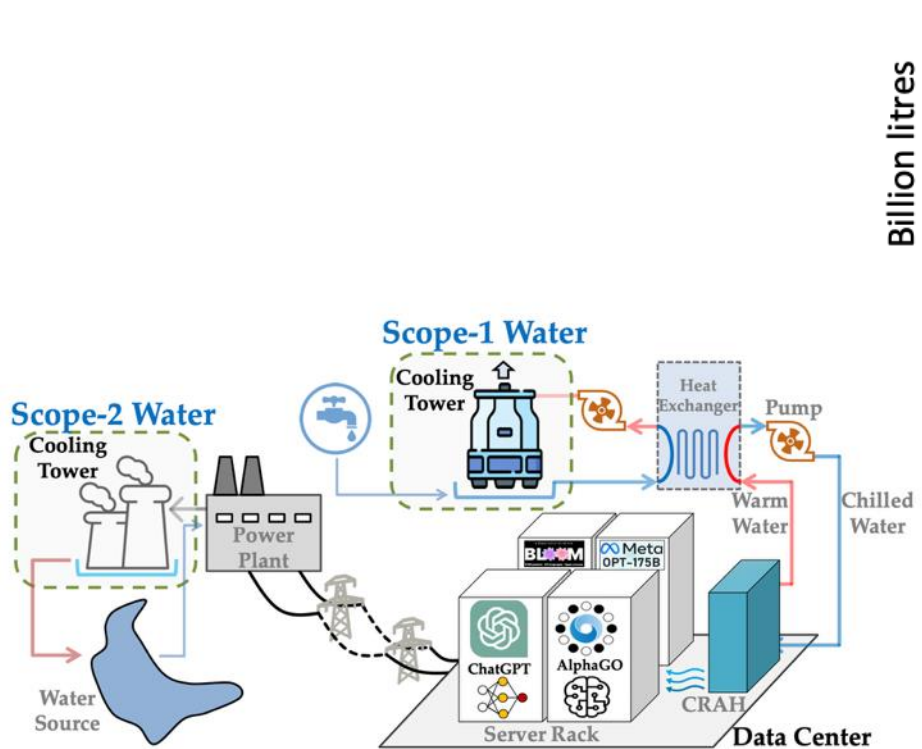
Alternative scenarios

 Algorithms improvement / frugality

 Lack of financing
 AI bubble burst
 Energy crunch

Water will logically follow the electricity trend

Massive adoption of evaporative cooling will add to the need of water for electricity generation



IEA. CC BY 4.0.

Water consumption more than doubles between 2023 and 2030

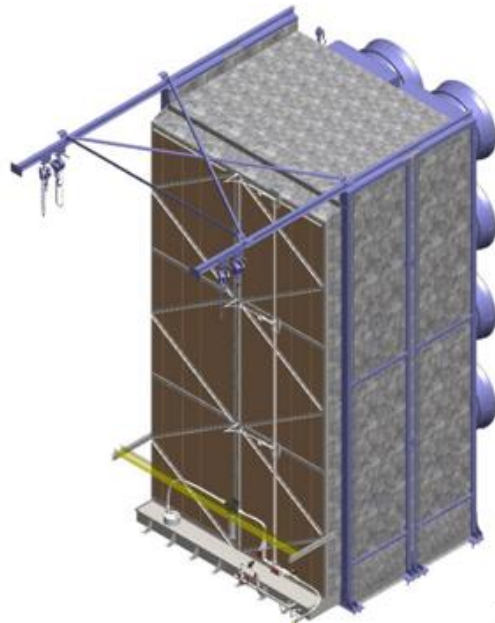
Source – « Energy and AI », IEA, 2025

Illustration : OVHcloud solution and its benefit on the WUE

Our proprietary evaporative cooling technology on dry coolers allows to keep the direct water usage low

High delta temperature and water profiles

Delta T at 20K (25-45°C)

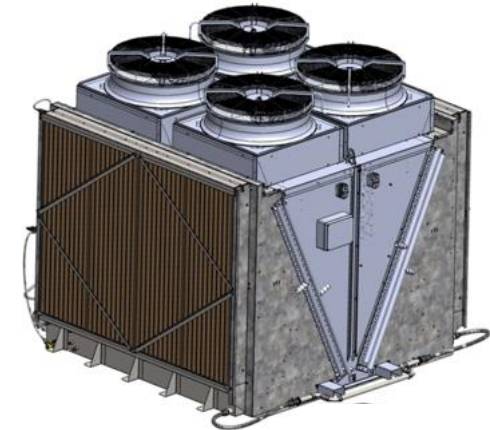


Worldwide Water Effectiveness

WUE* = 0.34 l/kWh (0.37 previous year)
(vs 1.00 to 1.50 l/kWh industry average)

Latest Designs Performance

WUE as low as 0.10 l/kWh



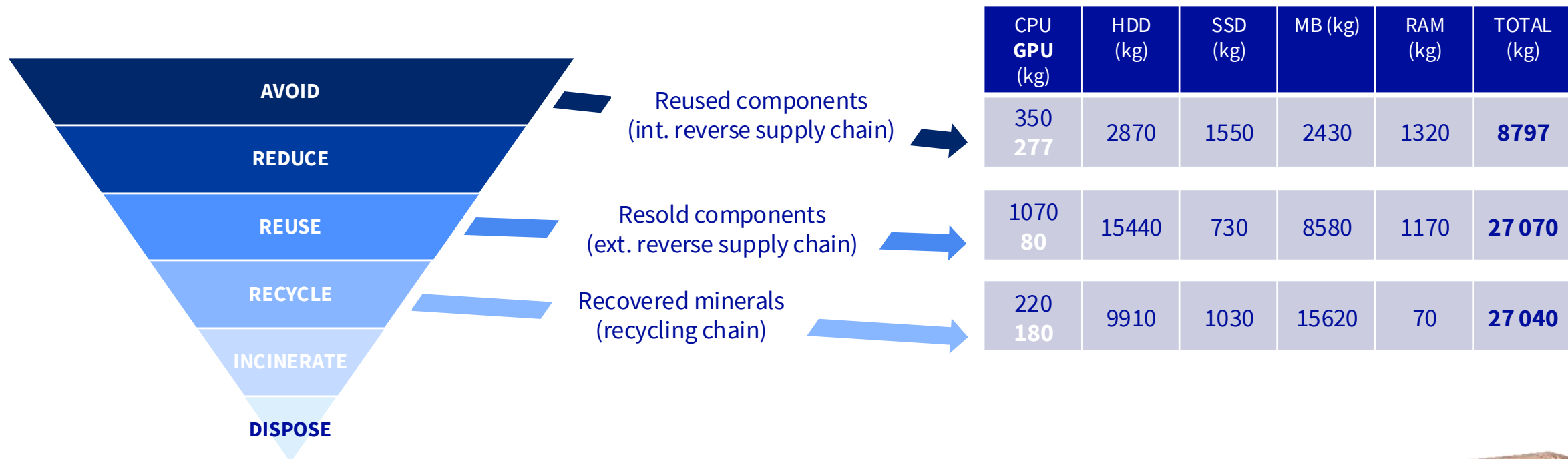
** ISO 30134 audited*

04 Beyond utilities



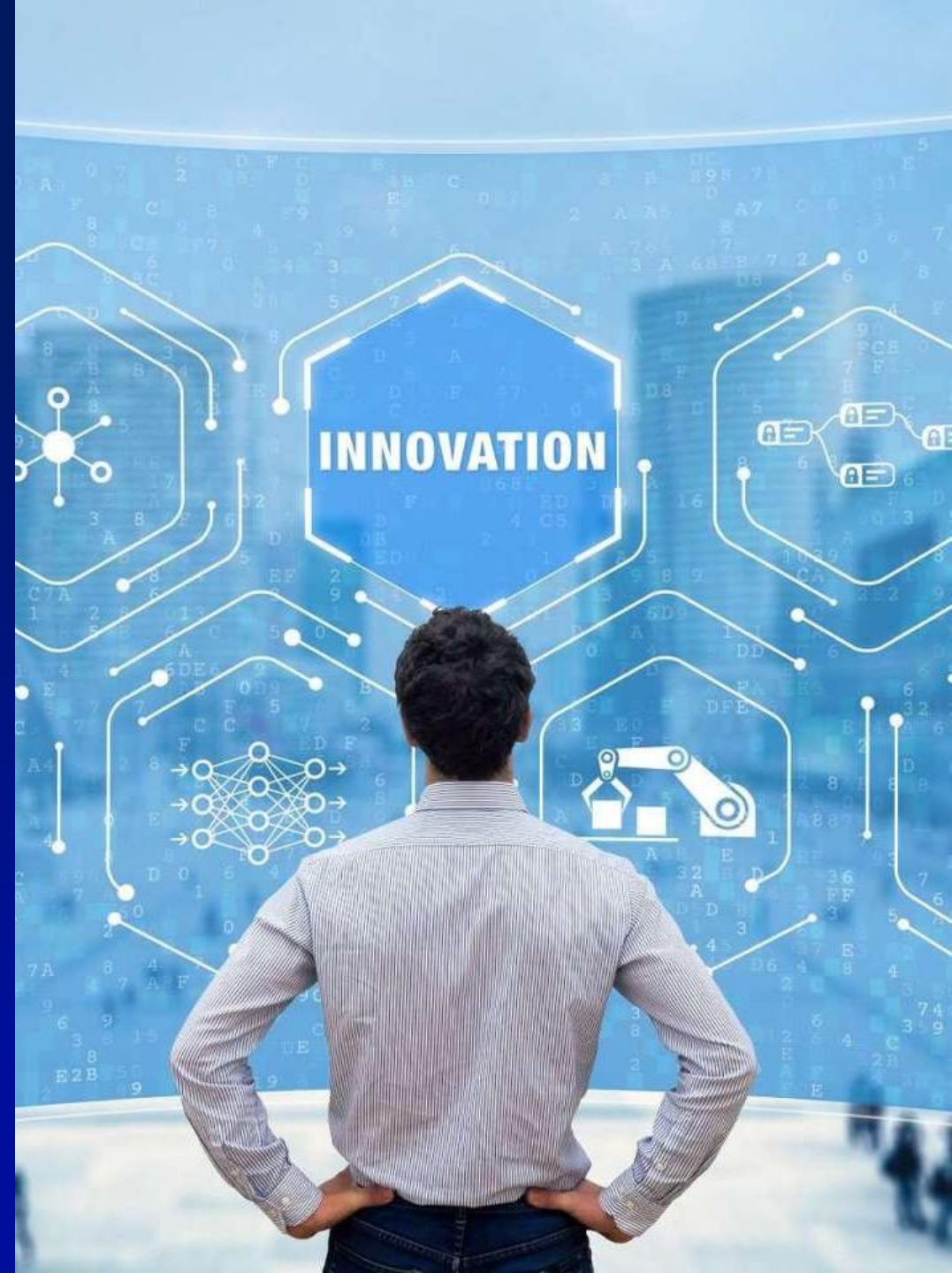
Comes last but is not the least : minerals recovery

GPUs are starting to pay off



05

Users and developers role



Choose the right GPU platform matching the target performance

Embodied emissions greatly vary from one reference to another

Typical “cradle to gate” values of new GPUs (to be amortized over 5 years)

| | |
|-------------------|--|
| ▶ Intel CPU range | 5 – 25 kgCO ₂ e |
| ▶ NVIDIA GPU L4 | 50 kgCO ₂ e |
| ▶ NVIDIA GPU L40s | 100 kgCO ₂ e |
| ▶ NVIDIA GPU A100 | 150 kgCO ₂ e |
| ▶ NVIDIA GPU H100 | 150 kgCO ₂ e (163 kgCO ₂ e**) |

Typical “cradle to gate” values of refurbished GPUs

| | |
|--------------------------------|-----------------------|
| ▶ NVIDIA Tesla V100 (2017) | 0 kgCO ₂ e |
| ▶ NVIDIA Quadro RTX5000 (2018) | 0 kgCO ₂ e |

*Source – Intel PCF / OVHcloud LCA

**Source – NVIDIA LCA

Choose the country based on your data location constraints

Usage emissions greatly vary from one country to another

NVIDIA H100



kCO2e/month (Location based)

- ▶ Manufacturing 73
- ▶ Operations 4
- ▶ Electricity 598

— - 85% →

NVIDIA H100



kCO2e/month (Location based)

- ▶ Manufacturing 73
- ▶ Operations 4
- ▶ Electricity 23

Choose the right AI instances in the portfolio of services

Pick up what you really need

AI end-points features to be looked at

- ▶ Quantisation optimisation (FP8 to FP4 = -50% in computing needs)
- ▶ Context caching (-30% to -40% in computing needs)
- ▶ Speculative decoding (small models first, then large models if results are not accurate enough)
- ▶ Model architecture change (model split in “n” expert models, the prompt is routed to the appropriate expert model)
- ▶ Batch processing (50k prompt treated asynchronously to optimize the GPU resources planning)
- ▶ Number of parameters reduction : 7 billions parameters models now as accurate a 100s billions parameters models 2 years ago