Not All Water Consumption Is Equal:

An AI, Datacenter, and Semiconductor Perspective

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Computing is thirsty

• The rise of Al and computing comes with a hidden cost: water consumption.

10~50 GPT-3 queries 500 mL water

A Google's datacenter 3.7 million m³ water

1,500



Intel

9.6 million m³ water

23,000





Water consumption is high across the entire computing stack.

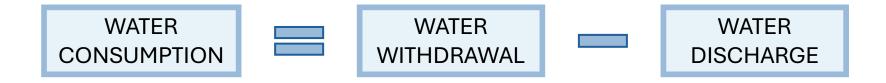
How much do we understand water consumption?

What is water consumption?

Water Withdrawal ≠ Water Consumption

Water Withdrawal: Total water taken from the environment

Water Consumption: Water removed from the environment



Water Consumption = On-site + Off-site

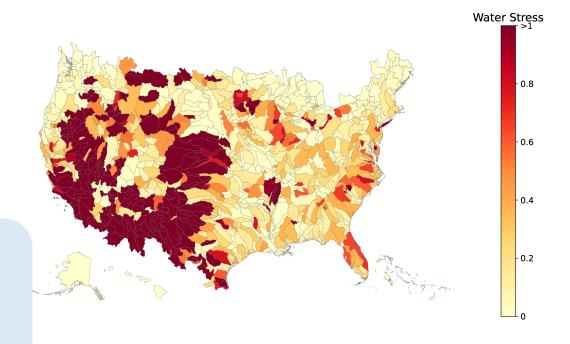
On-site consumption: Loss at facility (e.g., evaporation for colling in datacenters)

Off-site consumption: Indirect consumption (e.g., power generation)

Water consumption is location-sensitive

$$Water stress = \frac{Water Demand}{Water Availability} [1]$$

Spatially, same amount of water consumption can lead to greater environment burden in arid regions than in water-abundant regions.

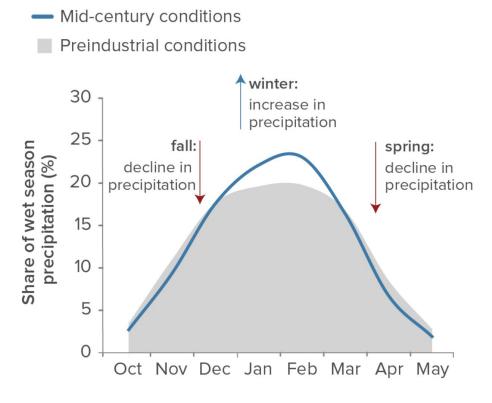


Water stress map of U.S.

Water consumption is time-sensitive

- Water availability fluctuates over seasons
- Climate change influences water in long-term

Temporally, water supply fluctuates over time. We must account for both the **location** and **time** when analyzing water impact.



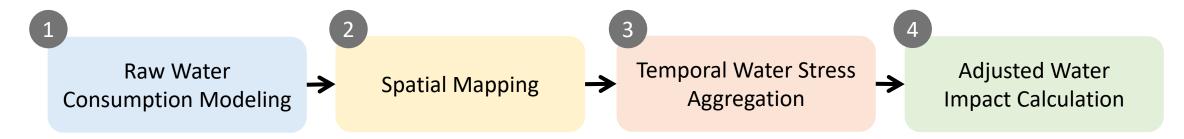
Seasonal Precipitation in California [2]

Our Contributions

- SCARF: Stress-Corrected Assessment of Water Resource Footprint, a general framework for water impact evaluation in computing.
 - Incorporate spatial and temporal variations of water stress to better evaluate water consumption impact.
 - Introduce Adjusted Water Impact (AWI)
 a unified metric combining water consumption with local water stress.
 - Conduct three case studies
 across AI, datacenters, and semiconductor fabs

SCARF

- SCARF: Stress-Corrected Assessment of Water Resource Footprint
- Insight: Incorporate spatial and temporal variations



Case Study I: AI – Large Language Models (LLMs)









Case Study I: LLM Serving

• Evaluate *per-request* Adjusted Water Impact (AWI) of 7B, 14B and 32B Qwen 2.5 model in different regions.

Low water stress region

The adjusted water impact of deploying LLMs is highly **location-sensitive**. Same workloads can have 1000× differences depending on where they are served.

Case Study I: LLM Serving

 Evaluate per-request Adjusted Water Impact (AWI) of 32B Qwen2.5 in different months but same regions (AZ vs WY).

Even in the same location, seasonal changes can significantly affect adjusted water impact. When we deploy matters—not just where.



Case Study II: Google Datacenters

Evaluate annual Adjusted Water Impact (AWI) of datacenters in different regions.

Low Stress

Water stress alone is not sufficient to capture true water impact. High water consumption in medium-stress regions can cause greater impact than expected.

High Stress

Case Study II: Google Datacenters

• Evaluate *annual* Adjusted Water Impact (AWI) of Google's datacenters with **different discount rates**.

Different planning perspectives lead to different sustainable decisions.



Case Study III: Intel Semiconductor Fabs

• Evaluate *annual* Adjusted Water Impact (AWI) of Intel's fabs in **different** regions.

Semiconductor fabs consume a large amount of water. Building fabs in water-scarce regions like Arizona causes higher environmental burdens.

Take Home Messages

SCARF quantifies water impact based on spatial and temporal variations in water stress.



Region matters.

Water stress level varies dramatically across regions.



Time matters.

Water stress shifts with seasons and long-term climate change.



Amount matters.

Large water consumption amplifies water impact even in low-stress area.



How we value the future matters.

Different priorities for future can reshape sustainability decisions.