Towards Environmentally *Sustainable* & *Equitable* Computing

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Partly based on joint work with Pengfei Li (UCR), Jianyi Yang (UCR), Mohammad A. Islam (UT Arlington) and Adam Wierman (Caltech).

UG RIVERSIDE



The demand for AI and computing is soaring!









The demand for AI and computing is soaring!

... and so is the sustainability concern!

At last — a computer program that can beat a champion Go player PAGE 48

ALL SYSTEMS GO

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Computing is resource-intensive and power-hungry





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Per IEA, data centers use about 300 TWh electricity in 2022, or 1-2% of the global electricity demand (0.2-0.4% global energy demand).







Electricity = Carbon





Electricity = Carbon + Water + Air & thermal pollution + ...





Electricity = Carbon + Water + Air & thermal pollution + ...





Carbon != Water

One footprint is "a complement to and not a substitute for" the other.

Kai Fang, Reinout Heijungs, Geert R. de Snoo, Theoretical exploration for the combination of the ecological, energy, carbon, and water footprints: Overview of a footprint family, Ecological Indicators, Volume 36, 2014,

Water withdrawal vs. water consumption for electricity

Take water from a source (e.g., groundwater)



Water withdrawal vs. water consumption for electricity

Take water from a source (e.g., groundwater)

1 kWh = roughly 44-100 L water withdrawal (excluding hydropower) 1 kWh = 3.14 L water consumption

Based on U.S. average water efficiency for electricity generation provided by EIA and WRI.







Data centers are guzzling water!



Figure 2. Multiple AI models are trained and/or deployed in the data center. Data center water footprint consists of two parts: on-site water and off-site water consumption

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Figure 2. Multiple AI models are trained and/or deployed in the data center. Data center water footprint



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Data centers are guzzling water!

- A large data center can consume millions of gallons of potable water each day for on-site cooling.
- Google's data center used 355 million gallons of water in The Dalles, OR, in 2021, 29% of the city's total water consumption

Google's annual water use in The Dalles, in gallons

What about Al's water usage (withdrawal)?

~6.6 billion cubic meters in 2027

Based on the projected AI GPU energy consumption of up to 134 TWh (de Vries, 2023), US average water withdrawal for electricity generation 44L/kWh, PUE=1.1, on-site water withdrawal 1L/kWh.

What about Al's water usage (consumption)?

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Based on the projected AI GPU energy consumption of up to 134 TWh (de Vries, 2023), US average water consumption for electricity generation 3.14 L/kWh, PUE=1.1, on-site water consumption 0.8L/kWh.

ChatGPT is already "drinking" a lot of water

Table 2: Estimate of GPT-3's average operational water footprint. "*" denotes data centers under construction as of July 2023, and the PUE and WUE values for these data centers are based on Microsoft's projection.

2		WITE	Electricity Water	Water for Training (million L)			Water for Inference (mL)			# of Inferences
Location	PUE	$(\mathbf{I} / \mathbf{W} \mathbf{h})$	Intensity	Onsite	Offsite	Total	Onsite	Offsite	Total	for 500ml
		(L/KVVII)	(L/kWh)	Water	Water	Water	Water	Water	Water	Water
US Average	1.170	0.550	3.142	0.708	4.731	5.439	2.200	14.704	16.904	29.6
Wyoming	1.125	0.230	2.574	0.296	3.727	4.023	0.920	11.583	12.503	40.0
Iowa	1.160	0.190	3.104	0.245	4.634	4.879	0.760	14.403	15.163	33.0
Arizona	1.223	2.240	4.959	2.883	7.805	10.688	8.960	24.259	33.219	15.1
Washington	1.156	1.090	9.501	1.403	14.136	15.539	4.360	43.934	48.294	10.4
Virginia	1.144	0.170	2.385	0.219	3.511	3.730	0.680	10.913	11.593	43.1
Texas	1.307	1.820	1.287	2.342	2.165	4.507	7.280	6.729	14.009	35.7
Singapore	1.358	2.060	1.199	2.651	2.096	4.747	8.240	6.513	14.753	33.9
Ireland	1.197	0.030	1.476	0.039	2.274	2.313	0.120	7.069	7.189	69.6
Netherlands	1.158	0.080	3.445	0.103	5.134	5.237	0.320	15.956	16.276	30.7
Sweden	1.172	0.160	6.019	0.206	9.079	9.284	0.640	28.216	28.856	17.3
Mexico*	1.120	0.056	5.300	0.072	7.639	7.711	0.224	23.742	23.966	20.9
Georgia*	1.120	0.060	2.309	0.077	3.328	3.406	0.240	10.345	10.585	47.2
Taiwan*	1.200	1.000	2.177	1.287	3.362	4.649	4.000	10.448	14.448	34.6
Australia*	1.120	0.012	4.259	0.015	6.138	6.154	0.048	19.078	19.126	26.1
India*	1.430	0.000	3.445	0.000	6.340	6.340	0.000	19.704	19.704	25.4
Indonesia*	1.320	1.900	2.271	2.445	3.858	6.304	7.600	11.992	19.592	25.5
Denmark*	1.160	0.010	3.180	0.013	4.747	4.760	0.040	14.754	14.794	33.8
Finland*	1.120	0.010	4.542	0.013	6.548	6.561	0.040	20.350	20.390	24.5

Estimates updated as of 09/2023 based on Microsoft's regional WUE data.

ChatGPT is already "drinking" a lot of water

ChatGPT needs about 500 ml of water for answering 10-50 questions.

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Water is a shared public good...

"Every drop matters." --- Meta

Water Positive by 2030!

Google Microsoft Meta amazon

Algorithmic challenges

"When" and "Where" matter a lot

- Outside weather condition
- Time-varying workloads

- Fuel mix for power generation
- Regional climate

Algorithmic challenges

"Follow the Sun" or "Unfollow the Sun"?

Water Efficient

Carbon Efficient

Learning-augmented algorithms

- Policy Prior," NeurIPS, 2023.
- Feedback Delay," NeurIPS, 2023.
- ICML, 2023.
- SIGMETRICS, 2022.

[NeurlPS'23] Jianyi Yang, Pengfei Li, Tongxin Li, Adam Wierman, and Shaolei Ren, "Anytime-Constrained Reinforcement Learning with [NeurlPS'23] Pengfei Li, Jianyi Yang, Adam Wierman, and Shaolei Ren, "Robust Learning for Smoothed Online Convex Optimization with [ICML'23] Pengfei Li, Jianyi Yang, and Shaolei Ren, "Learning for Edge-Weighted Online Bipartite Matching with Robustness Guarantees," [SIGMETRICS'22] Pengfei Li, Jianyi Yang, and Shaolei Ren, "Expert-Calibrated Learning for Online Optimization with Switching Costs,"

Environmental equity?

minimizing the total environmental cost

minimizing each region's environmental cost

Electricity = Carbon + Water + Air & thermal pollution + ...

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Computing's environmental inequity is emerging...

The uneven distribution of Al's The constant push for scale in artificial intelligence has led Big Tech firms to develop environmental cost is "historical hugely energy-intensive computational models that optimize for "accuracy" – through increasingly large datasets and computationally intensive model training – over more practices of settler colonialism efficient and sustainable alternatives. ① As we increasingly become locked into using Big-Tech infrastructures, we also become locked into their voracious appetite for and racial capitalism". resources, necessitating a life cycle analysis: the data centers needed for computationally intensive AI have high energy costs and carry a massive carbon footprint.② Computing technologies rely heavily on minerals that are procured under violent and exploitative conditions. ③ But these environmental harms are not evenly distributed; they disproportionately impact communities that are already marginalized, in a manner that reenacts historical practices of settler colonialism and racial

capitalism. (4)

In 2022, the United Nations Educational, **Scientific and Cultural Organization** (UNESCO) recommends that "Al should not be used" if it creates "disproportionate negative impacts on the environment".

Environmentally *equitable* geographical load balancing (eGLB)

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Equitably re-distribute the environmental cost across different regions!

cation Clients

Computing is "thirsty", just as we are!

ChatGPT needs 500 ml of water for answering 10-50 questions.

Estimates updated as of 09/2023 based on Microsoft's regional WUE data.

The water footprint is coming to the public...

Related papers

- [NeurIPS'23] Jianyi Yang, Pengfei Li, Tongxin Li, Adam Wierman, and <u>Shaolei Ren</u>, "Anytime-Constrained Reinforcement Learning with Policy Prior," NeurIPS, 2023.
- [NeurIPS'23] Pengfei Li, Jianyi Yang, Adam Wierman, and <u>Shaolei Ren</u>, "Robust Learning for Smoothed Online Convex Optimization with Feedback Delay," NeurIPS, 2023.
- [ICML'23] Pengfei Li, Jianyi Yang, and <u>Shaolei Ren</u>, "Learning for Edge-Weighted Online Bipartite Matching with Robustness Guarantees," ICML, 2023.
- [SIGMETRICS'22] Pengfei Li, Jianyi Yang, and <u>Shaolei Ren</u>, "Expert-Calibrated Learning for Online Optimization with Switching Costs," SIGMETRICS, 2022.
- [Preprint'23] Pengfei Li, Jianyi Yang, Adam Wierman, and <u>Shaolei Ren</u>, "Towards Environmentally Equitable AI via Geographical Load Balancing," arXiv, 2023.
- [Preprint'23] Pengfei Li, Jianyi Yang, Mohammad A. Islam, <u>Shaolei Ren</u>, "Making AI Less 'Thirsty': Uncovering and Addressing the Secret Water Footprint of AI Models," arXiv, 2023.

Thanks!

