

ORCA

Outcome-aware
Regional
Compute
Allocation



Sustainability-Aware Workload Shifting Beyond Carbon Intensity

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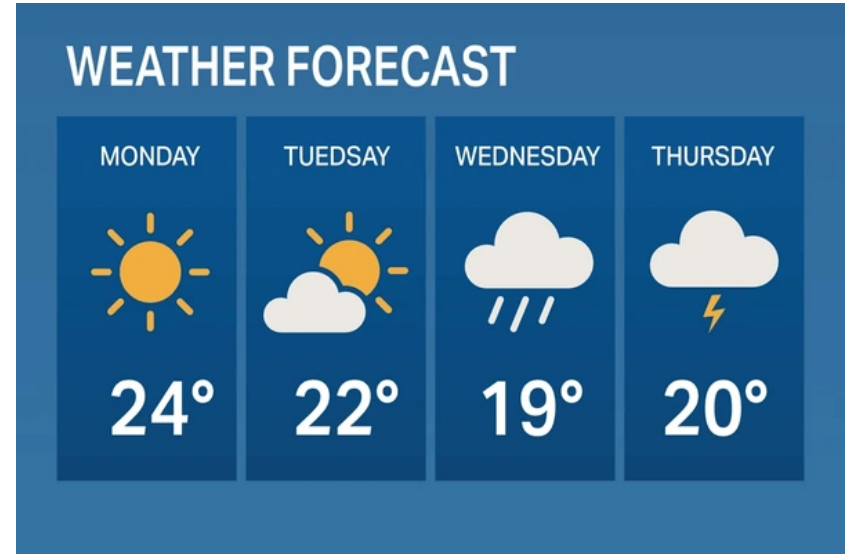
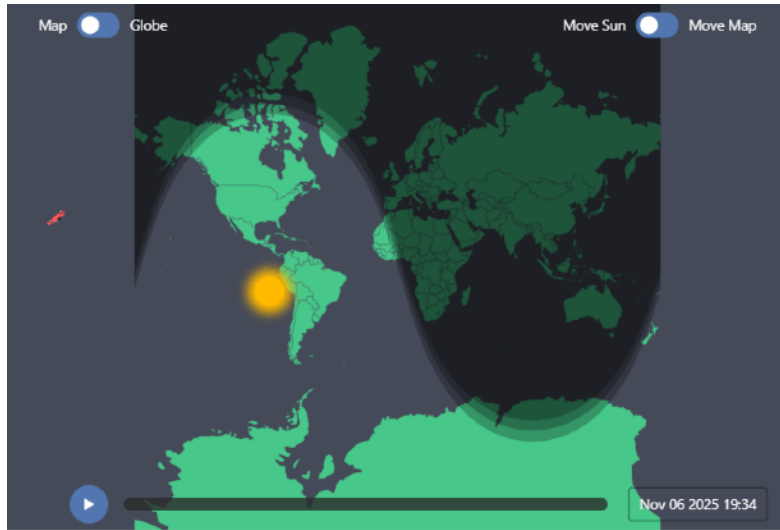
Sustainability-Aware Workload Shifting Beyond Carbon Intensity

RQ1: How can workload shifting consider local impacts in addition to CO₂?

RQ2: When does CO₂-optimal shifting increase local burdens?

RQ3: How does zone-aware workload placement compare to CO₂-only workload placement?

Location and Time shifting



Not only CO2 is harming people



<https://arstechnica.com/tech-policy/2025/04/elon-musks-xai-accused-of-lying-to-black-communities-about-harmful-pollution/>

<https://datacenters.atmeta.com/2025/10/hello-el-paso/>

<https://datacentremagazine.com/news/how-google-is-lowering-data-centre-emissions-amid-ai-growth>

Google is just great



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Google is just great, thx Max



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The AI boom is built on the back of the down trodden

Environment Science Global development Football Tech Business Startups

NAACP lawsuit accuses Elon Musk's xAI of polluting Black neighborhoods near Memphis

Suit alleges the billionaire's AI company is illegally spewing toxic pollutants from its datacenter in the Memphis area

CLOSE ALL TABS

How the AI Data Center Boom Impacts Black Communities



LISTEN

By Morgan Sung, Maya Cueva, Chris Egusa, Chris Hambrick Feb 4, 2026 Updated Mar 10, 2026 Save Article



Tech Policy Press

Data Center Boom Risks Health of Already Vulnerable Communities

This analysis aims to build awareness of the environmental and health risks that generative AI's energy expansion poses to marginalized communities.

12.06.2025

<https://naacp.org/articles/elon-musks-xai-threatened-lawsuit-over-air-pollution-memphis-data-center-filed-behalf>

<https://www.canarymedia.com/articles/fossil-fuels/how-the-data-center-boom-could-harm-black-communities>

What are other impacts

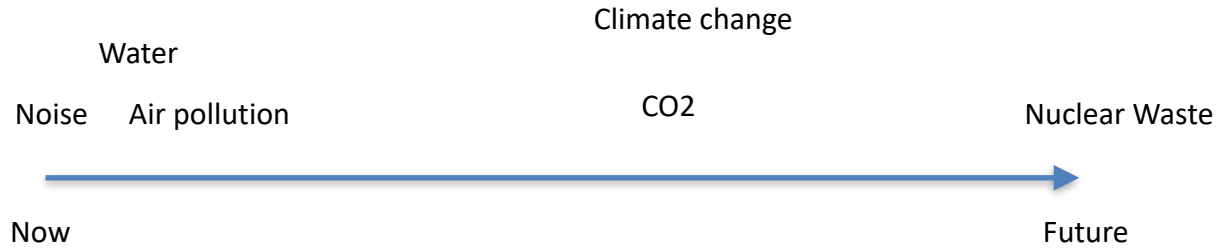
1. Embodied Emissions
 1. Metals and other Elements
 2. Transportation
2. Electricity ✓
3. Water ✓
4. Land usage
5. Rare earths
6. Waste management
7. Air Pollution ✓
8. Human

New factors emerging:

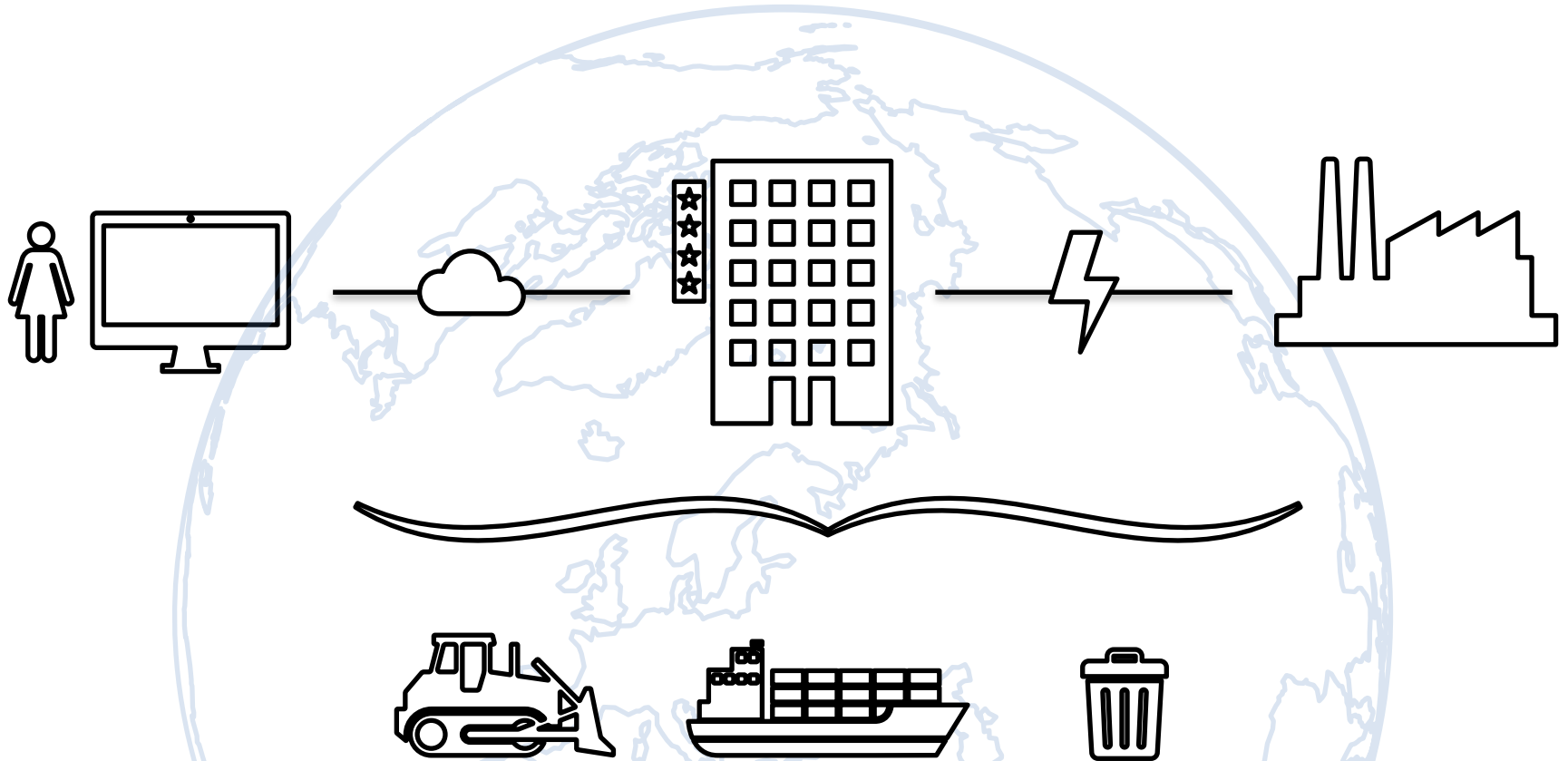
- Noise
- Light
- Refrigerant leakage
- Metal for construction
- Social / Society

CO2 vs the others

How should we rate which one is more important?



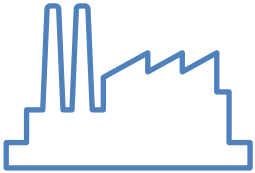
Actors



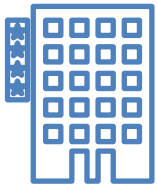
New Concept => Impact Zones



Global Zone: CO₂ -eq (g/kWh)

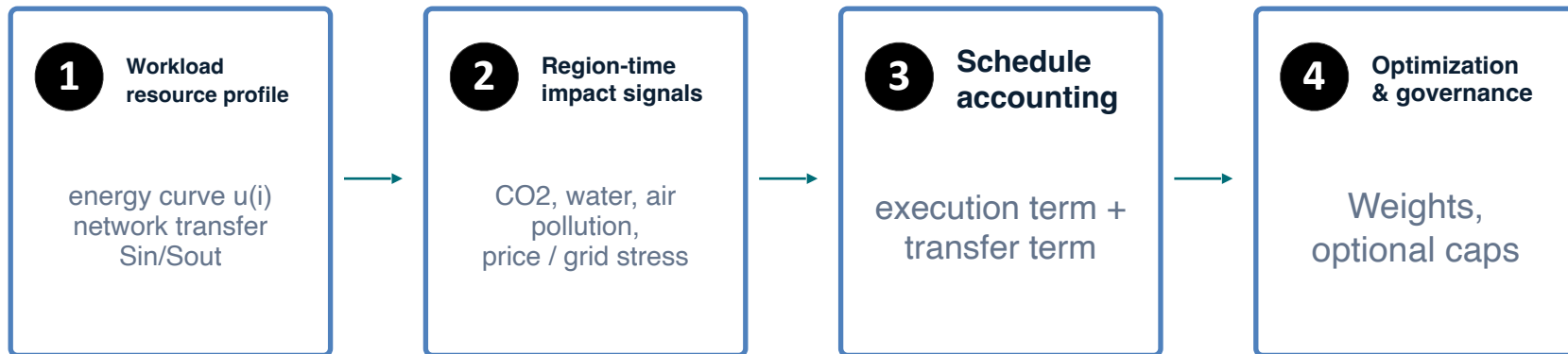


Energy Production Zone: SO₂ -eq (g/kWh) CO (g/kWh)



Data-Centre Zone: NO₂ (μg/m³) PM_{2.5} (μg/m³) Price (EUR) Water

Orca decision model: from workload traces to auditable placement



Decision variable: choose one data-centre d and one feasible start time τ for a non-migratory workload.

Formalization:

Schedule-induced burden per dimension

$$I_e(d, \tau) = \underbrace{\sum_{i=0}^{n-1} u(i) \cdot g_{d,e}(\tau + i\Delta)}_{\text{execution term}} + \underbrace{T_e(o, d)}_{\text{transfer term}}.$$

Weighted objective over normalized loads

$$J(d, \tau) = \sum_{e \in \mathcal{E}} \omega_e \frac{I_e(d, \tau)}{s_e}$$

Constraint layer: $I_e(d, \tau) \leq \text{Cap}_e(\text{zone}(d,e))$ for each dimension e

Caps prevent global optimization from violating local burden limits.

Experiment

Locations (all northern hemisphere)

DE	Europe (Frankfurt)
JP	Asia-Pacific (Tokyo)
CA	U.S. West (Northern California)

2 Dates:

- 01.07.2025 (summer)
- 01.01.2026 (winter)

Workload - Django unit tests

$H = 883.26 \text{ s}$

$E_{\text{exec}} = 6905.28 \text{ mWh} = 0.00690528 \text{ kWh}$

Very Uniform workload

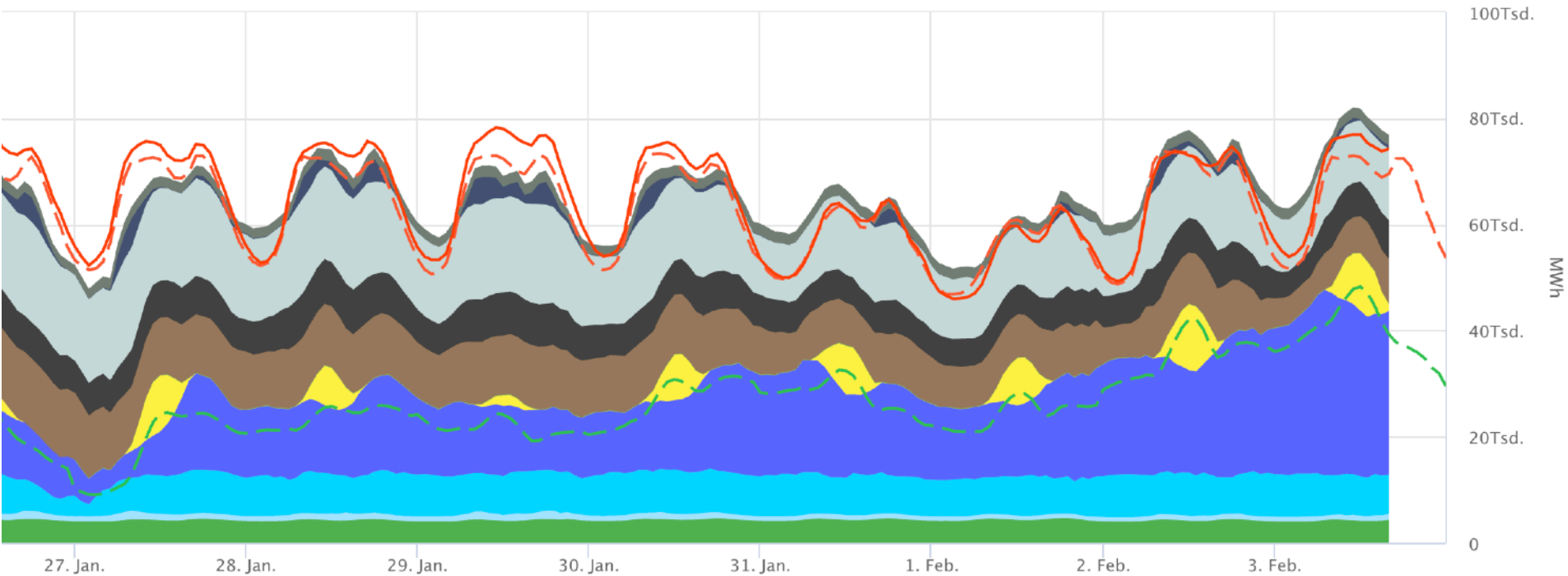
$S_{\text{in}} = 51.70 \text{ MB}$

$S_{\text{out}} = 0 \text{ MB}$

$\approx 0.46 \text{ gCO}_2\text{e}$

Other impacts are for future work

Grids are really complex

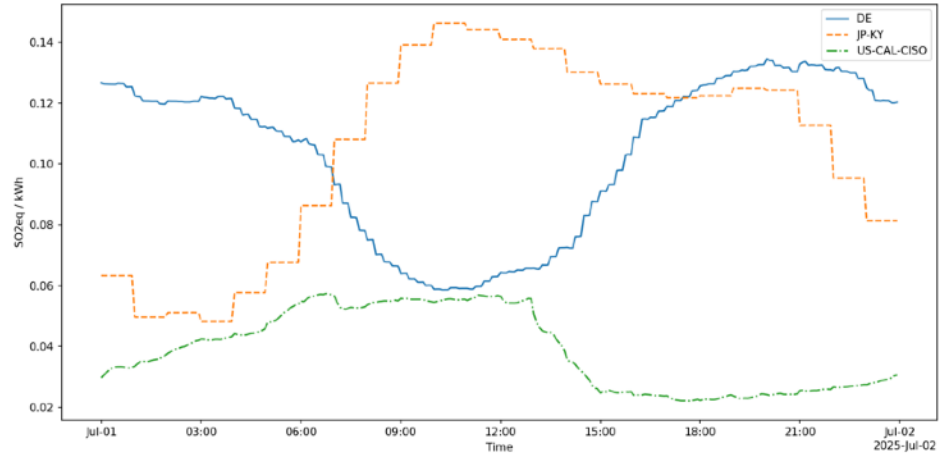
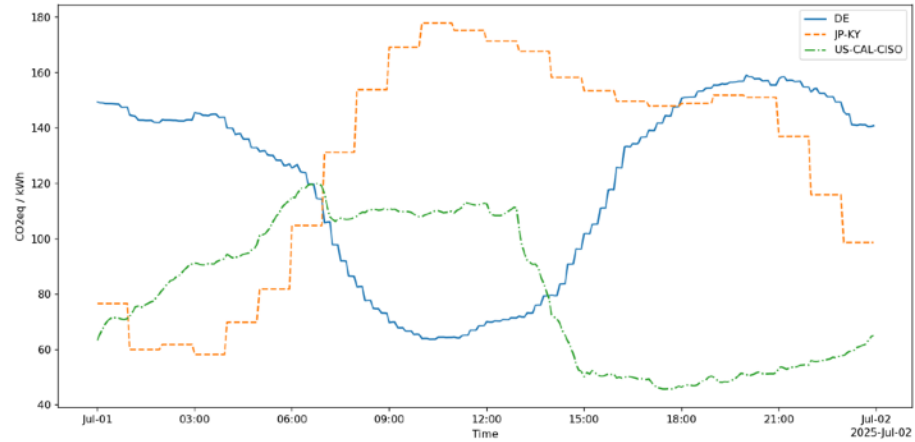


There is not only CO2

TABLE II: Emissions in g per kWh electricity

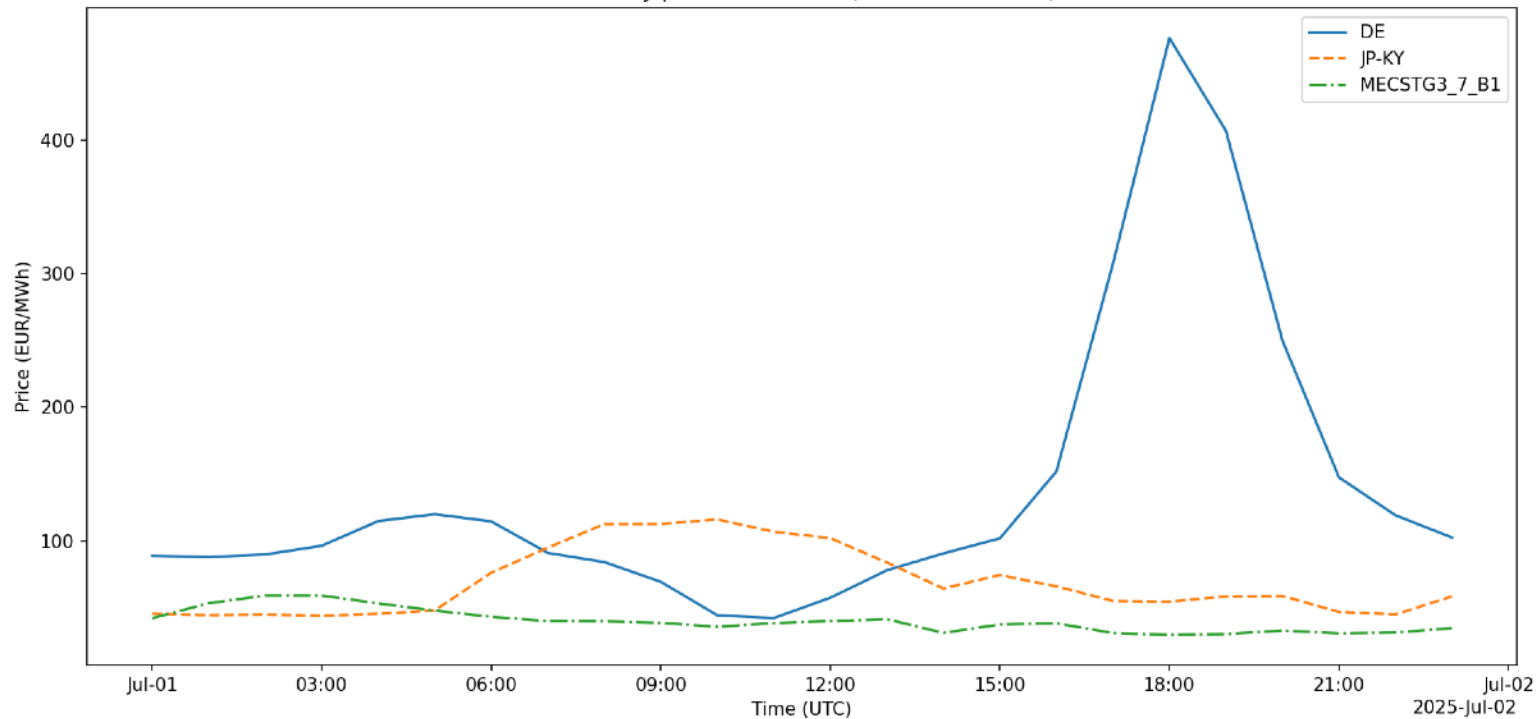
Energy Source	CO ₂ -eq	CO ₂	CH ₄	N ₂ O	SO ₂ -eq	SO ₂	NO _x	CO	NM-VOC
nuclear	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
geothermal	26.68	24.854	0.05	0.002	0.037	0.015	0.031	0.024	0.002
biomass	12.871	11.989	0.024	0.001	0.018	0.007	0.015	0.012	0.001
coal	398.36	394.95	0.002	0.013	0.387	0.196	0.279	0.114	0.005
wind	0.87	0.81	0.002	0.0	0.001	0.0	0.001	0.001	0.0
solar	0.44	0.41	0.001	0.0	0.001	0.0	0.001	0.0	0.0
hydro	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
hydro discharge	8.89	8.28	0.017	0.001	0.012	0.005	0.01	0.008	0.001
gas	212.3	202.77	0.295	0.005	0.095	0.001	0.136	0.05	0.005
oil	268.78	266.47	0.023	0.006	0.505	0.169	0.482	0.198	0.0

CO2eq vs SO2eq

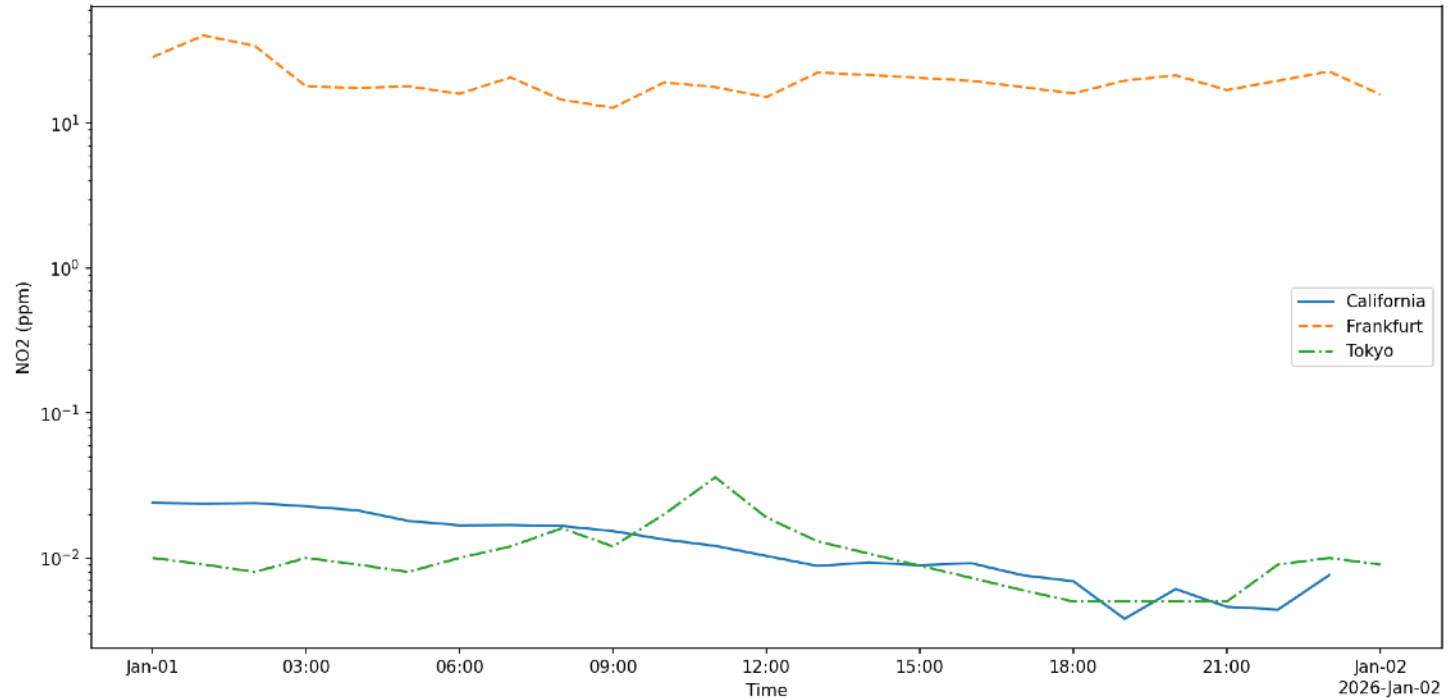


Electricity Price

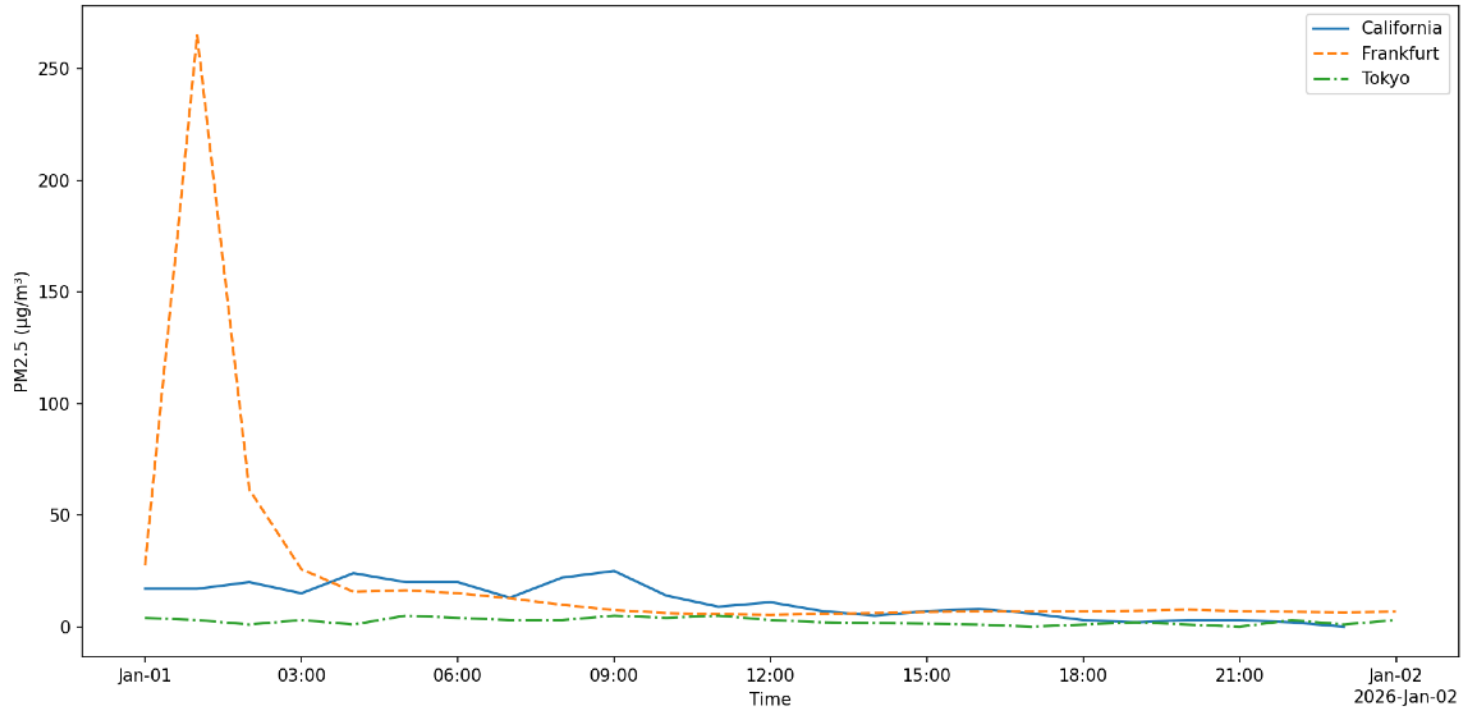
Electricity price time series (converted to EUR)



Air NO2 - OpenAQ



Air PM2.5 - OpenAQ



Water

Europe (Frankfurt)	0.01
Asia-Pacific (Tokyo)	0.91
U.S. West (Northern California)	0.51

TABLE III: Water scarcity risk levels by region and date.

Date	Frankfurt	Tokyo	Northern California
01.01.2026	Medium-High	Medium-High	Low
01.07.2025	High	Medium-High	Medium-High

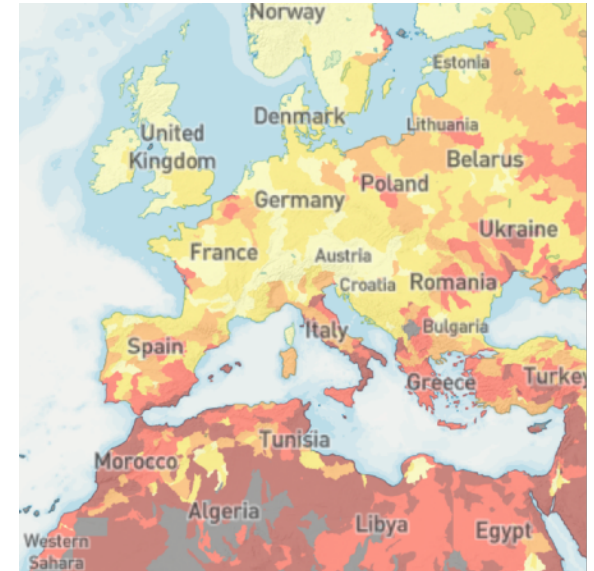


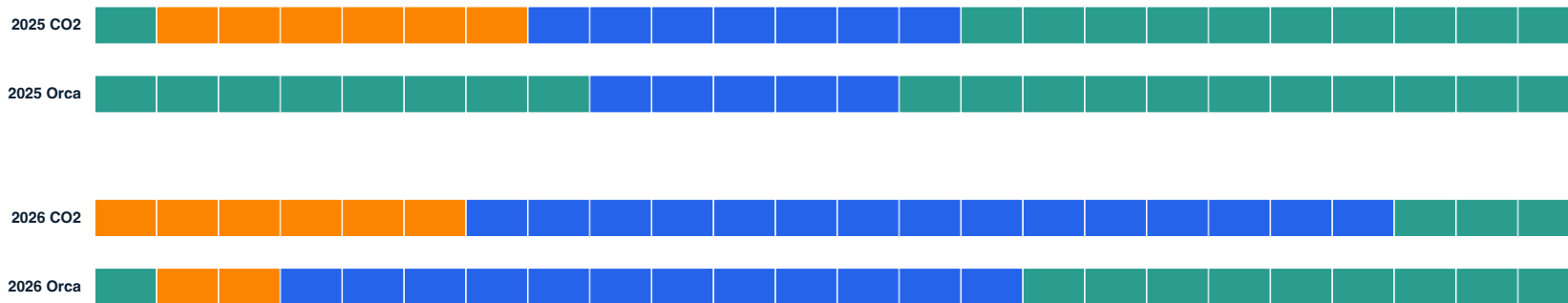
TABLE V: Hourly intensities by region with lowest values highlighted

Time (h)	Global Zone			Energy Production Zone						Data-Centre Zone											
	CO ₂ -eq (g/kWh)			SO ₂ -eq (g/kWh)			CO (g/kWh)			NO ₂ ($\mu\text{g}/\text{m}^3$)			PM _{2.5} ($\mu\text{g}/\text{m}^3$)			Price (EUR)			Water		
	CA	DE	JP	CA	DE	JP	CA	DE	JP	CA	DE	JP	CA	DE	JP	CA	DE	JP	CA	DE	JP
2025																					
00:00	69.57	148.57	76.64	0.03	0.13	0.06	0.02	0.04	0.02	0.01	15.10	0.01	1.00	5.86	12.00	42.37	89.10	45.64	1.53	0.04	2.73
01:00	76.20	143.08	60.01	0.04	0.12	0.05	0.02	0.04	0.02	0.01	11.13	0.01	5.00	4.58	13.00	53.52	88.08	44.69	1.53	0.04	2.73
02:00	86.44	142.85	61.83	0.04	0.12	0.05	0.02	0.04	0.02	0.01	9.22	0.01	3.00	3.93	13.00	59.27	90.01	45.17	1.53	0.04	2.73
03:00	91.35	144.72	58.24	0.04	0.12	0.05	0.02	0.04	0.02	0.01	10.80	0.01	4.00	4.12	15.00	59.27	96.50	44.25	1.53	0.04	2.73
04:00	94.65	136.68	69.78	0.04	0.12	0.06	0.02	0.04	0.02	0.01	8.00	0.01	5.00	3.99	13.00	53.40	114.97	45.64	1.53	0.04	2.73
05:00	107.05	129.23	81.89	0.05	0.11	0.07	0.03	0.04	0.02	0.01	15.18	0.01	6.00	4.38	14.00	48.04	120.11	48.13	1.53	0.04	2.73
06:00	118.23	121.14	104.73	0.06	0.10	0.09	0.03	0.03	0.03	0.01	27.11	0.01	5.00	4.87	15.00	43.40	114.72	76.38	1.53	0.04	2.73
07:00	108.37	95.54	131.24	0.05	0.09	0.11	0.03	0.03	0.04	0.01	37.98	0.01	6.00	5.60	11.00	40.19	91.19	95.08	1.53	0.04	2.73
08:00	110.02	77.10	153.89	0.05	0.07	0.13	0.03	0.02	0.04	0.01	48.72	0.01	10.00	5.85	11.00	39.97	84.31	112.77	1.53	0.04	2.73
09:00	109.71	67.44	169.15	0.06	0.06	0.14	0.03	0.02	0.05	0.01	39.01	0.01	8.00	5.98	10.00	38.84	69.56	112.86	1.53	0.04	2.73
10:00	109.45	64.11	177.95	0.06	0.06	0.15	0.03	0.02	0.05	0.01	41.34	0.02	7.00	6.79	13.00	35.95	44.54	116.21	1.53	0.04	2.73
11:00	111.89	66.23	175.39	0.06	0.06	0.14	0.03	0.02	0.05	0.01	35.91	0.02	5.00	6.45	13.00	38.67	42.47	107.05	1.53	0.04	2.73
12:00	109.79	70.59	171.40	0.05	0.06	0.14	0.03	0.02	0.05	N/A	32.35	0.01	3.00	6.36	9.00	40.15	57.65	102.16	1.53	0.04	2.73
13:00	90.07	75.11	167.73	0.04	0.07	0.14	0.02	0.02	0.05	0.01	N/A	0.01	7.00	N/A	9.00	41.61	78.14	84.02	1.53	0.04	2.73
14:00	63.18	87.62	158.29	0.03	0.08	0.13	0.02	0.03	0.05	0.01	33.56	N/A	8.00	7.02	N/A	31.38	90.80	64.58	1.53	0.04	2.73
15:00	51.07	108.98	153.54	0.03	0.10	0.13	0.01	0.03	0.04	0.01	25.50	N/A	10.00	5.42	N/A	37.62	102.02	74.63	1.53	0.04	2.73
16:00	49.72	132.52	149.74	0.02	0.11	0.12	0.01	0.04	0.04	0.01	25.52	0.02	6.00	5.29	11.00	38.54	152.08	66.21	1.53	0.04	2.73
17:00	46.53	143.27	148.05	0.02	0.12	0.12	0.01	0.04	0.04	0.01	35.31	0.02	6.00	6.24	10.00	31.24	307.27	55.22	1.53	0.04	2.73
18:00	47.95	152.48	148.86	0.02	0.13	0.12	0.01	0.04	0.04	0.01	39.53	0.01	8.00	6.23	9.00	29.92	476.19	54.71	1.53	0.04	2.73
19:00	49.36	156.26	151.91	0.02	0.13	0.12	0.01	0.05	0.04	0.01	37.98	0.01	6.00	6.71	9.00	30.50	406.81	58.62	1.53	0.04	2.73
20:00	51.24	157.31	151.07	0.02	0.13	0.12	0.01	0.05	0.04	0.01	56.31	0.01	7.00	8.36	12.00	33.12	250.01	58.98	1.53	0.04	2.73
21:00	53.79	156.82	137.01	0.03	0.13	0.11	0.01	0.04	0.04	0.01	72.99	0.01	6.00	12.26	11.00	31.02	147.65	47.07	1.53	0.04	2.73
22:00	56.46	151.80	115.87	0.03	0.13	0.10	0.01	0.04	0.03	0.01	68.92	0.01	11.00	11.89	12.00	31.59	119.29	45.17	1.53	0.04	2.73
23:00	61.06	142.09	98.69	0.03	0.12	0.08	0.01	0.04	0.03	0.01	40.16	0.02	9.00	8.84	12.00	35.00	102.70	58.65	1.53	0.04	2.73
2026																					
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01:00	93.33	72.62	21.13	0.05	0.06	0.02	0.02	0.02	0.01	0.02	40.24	0.01	17.00	265.15	3.00	38.50	19.57	43.98	0.51	0.03	2.73

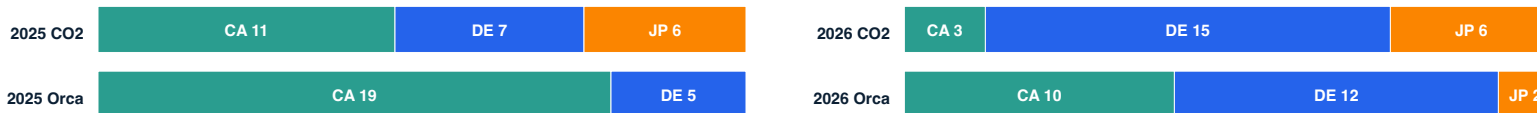
Result: Orca chooses different locations than CO2-only routing

■ CA - Northern California ■ DE - Frankfurt ■ JP - Tokyo

Hourly selected data-centre (UTC)



Regional allocation counts (hours/day)



What we answered

RQ1: How can workload shifting consider local impacts in addition to CO₂?

=> Zones

RQ2: When does CO₂-optimal shifting increase local burdens?

=> Heavily depends on the local setting but can be modeled

RQ3: How does zone-aware workload placement compare to CO₂-only workload placement?

=> Heavily depends on the weighting between impacts but there is a difference

Limitations: the model is exact about accounting, not omniscient about signals

■ Feasibility boundary

Assumes legal, latency, deployment and service constraints are already applied.

■ Non-migratory jobs

No preemption, checkpoint/restart, or live migration in the base model.

■ Transfer simplification

Per-MB proxy ignores path, caching, congestion, and transfer time.

■ Average vs marginal mix

Average generation mix is illustrative; marginal damage signals are better when available.

■ Data quality

Missing, stale, or sparse local signals can bias normalized objectives.

■ Embodied impacts

Omitted from example; can be added via amortized hardware burden terms.

The case study is a transparent demonstration of a decision model, not a universal operational recommendation for these three regions.

Conclusion

1. We can show that there are other values than CO2 that can be relevant
2. Emissions are not always global but increase burdens locally
3. Zones can model this
4. When taking into account other impacts routing decisions change

Next Paper: Why are people not migrating workloads?

I will need your help with this one

Questions?

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