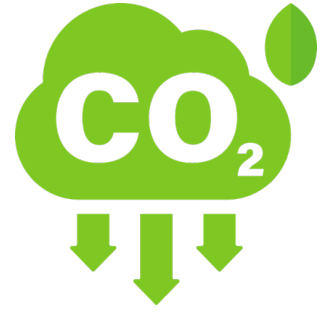




The Hidden Carbon Footprint of Serverless Computing



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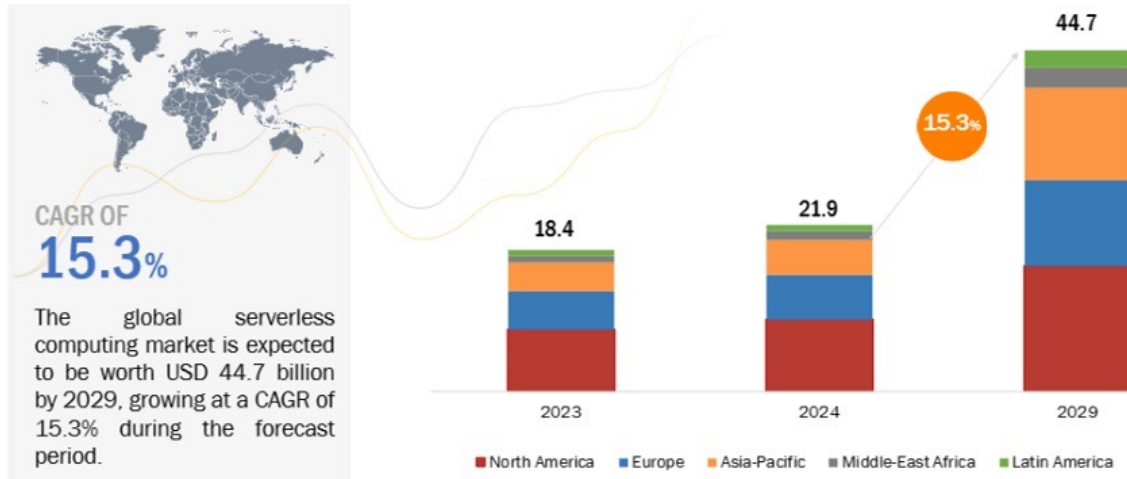
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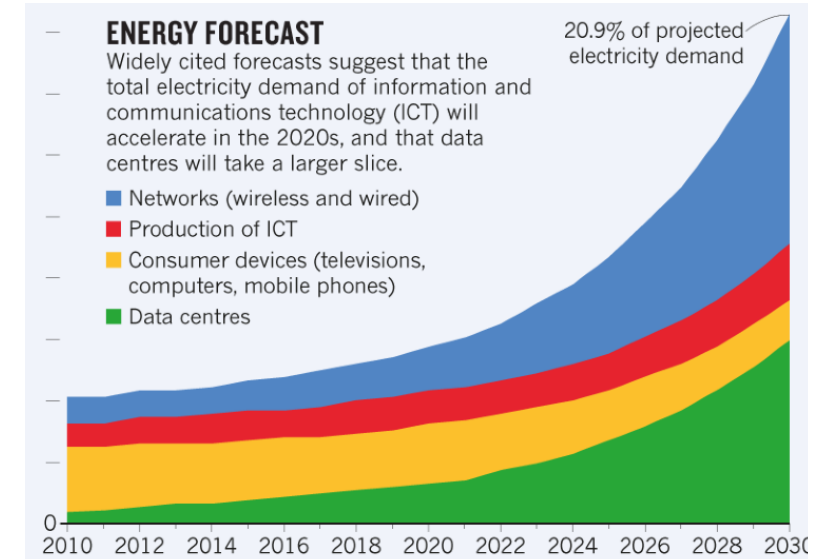


Why are we interested in this problem?



* sourced from <https://www.datadoghq.com/state-of-serverless/>

More than 70% of users of production cloud platforms like AWS, using one or more serverless solutions



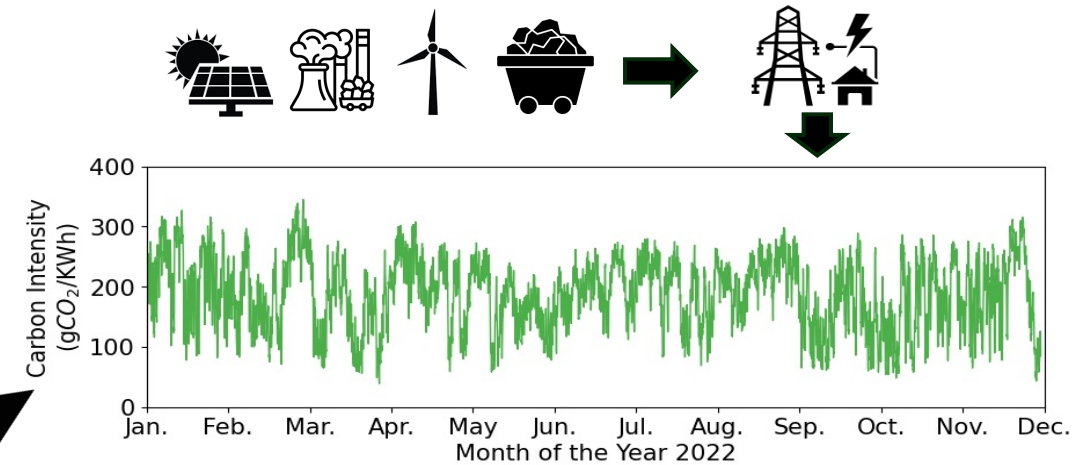
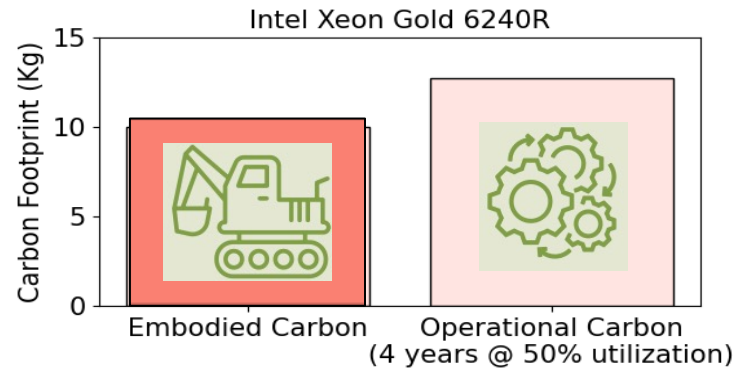
* sourced from <https://www.nature.com/articles/d41586-018-06610-y>

The energy demand, and the carbon footprint of datacenters is projected to grow at a rapid rate



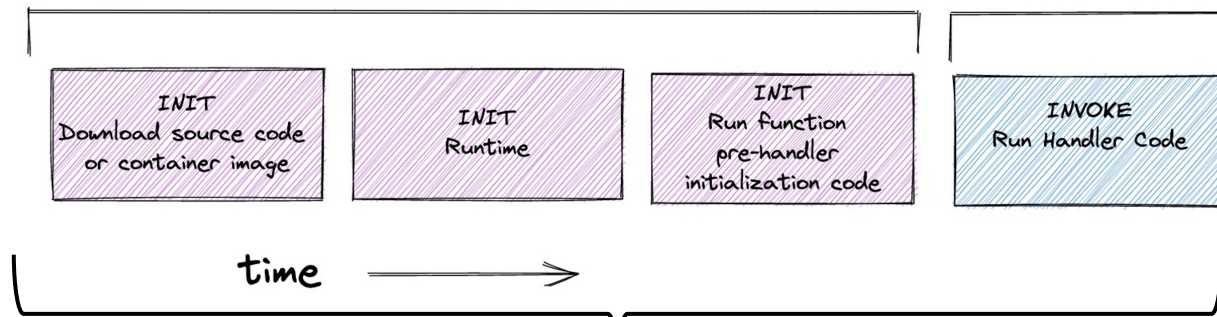
Why is carbon accounting for serverless challenging?

Embodied carbon
changes with
hardware and
operational carbon
changes with
carbon intensity



Cold Start

Warm Start



Execution



Keep-alive carbon

Execution carbon

Sharing resources among
functions during execution and
keep-alive make carbon
accounting challenging

Accounting for serverless carbon footprint is challenging and methodologically error-prone because of the unique keep-alive period aspects of serverless functions

Keep-alive carbon has operational and embodied components

$$C_{f_{\text{keep-alive}}} = \frac{E_{\text{DRAM}}}{L_{\text{DRAM}}} * T_{\text{keep-alive}} + CI * J_{\text{DRAM}}$$

Choice A

Diagram labels and arrows:

- DRAM Embodied Carbon** (blue arrow pointing down to E_{DRAM})
- Keep-alive Time** (blue arrow pointing down to $T_{\text{keep-alive}}$)
- Carbon Intensity** (blue arrow pointing down to CI)
- DRAM Lifetime** (blue arrow pointing up to L_{DRAM})
- Embodied Carbon** (red arrow pointing up to the pink box containing the first term)
- DRAM Energy** (blue arrow pointing up to J_{DRAM})
- Operational Carbon** (green arrow pointing left to the second term)

DRAM's embodied carbon has a significant environmental impact. This assumes that only one function is running/ kept alive in a server.

Keeping functions alive on harvested memory improves performance and should be included in carbon calculation

Fraction of memory consumed by function

$$C_{f_{\text{keep-alive}}} = \underbrace{\frac{E_{\text{DRAM}} * \Delta f_{\text{DRAM}}}{L_{\text{DRAM}}} * T_{\text{keep-alive}}}_{\text{Embodied Carbon}} + \underbrace{\text{CI} * J_{\text{DRAM}} * \Delta f_{\text{DRAM}}}_{\text{Operational Carbon}}$$

Choice B

But CPU cores should remain reserved during keep-alive so that functions can run when invoked.
Where is their carbon?

Reservation of CPU cores during keep-alive period consumes embodied carbon

$$C_{f_{\text{keep-alive}}} = \left(\frac{E_{\text{DRAM}} * \Delta f_{\text{DRAM}}}{L_{\text{DRAM}}} + \frac{E_{\text{CPU}} * \frac{n_f}{N_{\text{CPU}}}}{L_{\text{CPU}}} \right) * T_{\text{keep-alive}} + \text{CI} * J_{\text{DRAM}} * \Delta f_{\text{DRAM}}$$

Choice C

Embodied DRAM

Embodied CPU

Fraction of cores reserved

Operational DRAM

The cores still consume idle power even when they are not actively executing functions during the keep-alive period

Keep-alive carbon should include the operational and embodied carbon of DRAM and CPU

$$C_{f_{\text{keep-alive}}} = \left(\frac{E_{\text{DRAM}} * \Delta f_{\text{DRAM}}}{L_{\text{DRAM}}} + \frac{E_{\text{CPU}} * \frac{n_f}{N_{\text{CPU}}}}{L_{\text{CPU}}} \right) * T_{\text{keep-alive}} +$$

Embodied DRAM

Embodied CPU

Choice D

$$CI * \left(J_{\text{DRAM}} * \Delta f_{\text{DRAM}} + J_{\text{CPU}} * \lambda_{\text{IDLE}} * \frac{n_f}{N_{\text{CPU}}} \right)$$

Operational DRAM

Operational CPU

Idle power factor

Since CPU cores are indirectly consumed, their carbon accounting is harder than DRAM due to varying implicit reservations across providers

Functions lack fixed hardware, making it harder to estimate carbon impact across diverse systems

$$C_{f_{\text{execution}}} = \left(\frac{E_{\text{DRAM}} * \Delta f_{\text{DRAM}}}{L_{\text{DRAM}}} + \frac{E_{\text{CPU}} * \frac{n_f}{N_{\text{CPU}}}}{L_{\text{CPU}}} + \left(\frac{E_{\text{ST}} * \frac{S_f}{S_{\text{ST}}}}{L_{\text{ST}}} \right) * T_{\text{exe}} + \text{CI} * \left(J_{\text{DRAM}} * \Delta f_{\text{DRAM}} + J_{\text{CPU}} * \frac{n_f}{N_{\text{CPU}}} + J_{\text{ST}} * \frac{S_f}{S_{\text{ST}}} \right) \right)$$

Embodied DRAM

Embodied CPU

Operational DRAM

Operational Storage

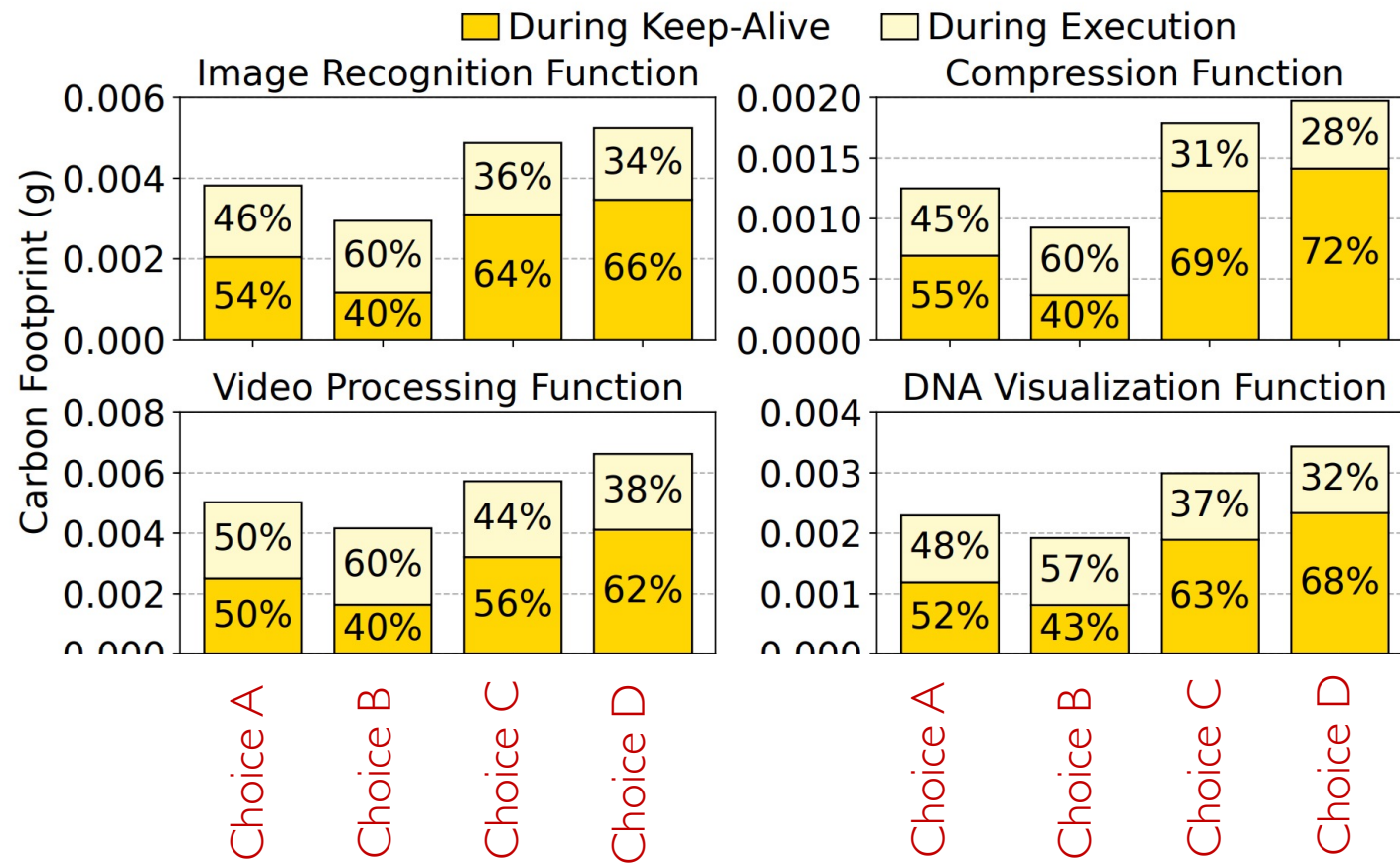
Operational CPU

Embodied Storage

The short runtime and low resource needs of serverless functions make them ideal for co-location, but this complicates accurate carbon accounting, requiring precise energy measurement

What are the implications of different methodological choices and scheduling on the serverless carbon footprint?

Keep-alive outweigh execution in carbon due to brief executions and extended keep-alive periods



Choice A: DRAM embodied and operational for keep-alive carbon

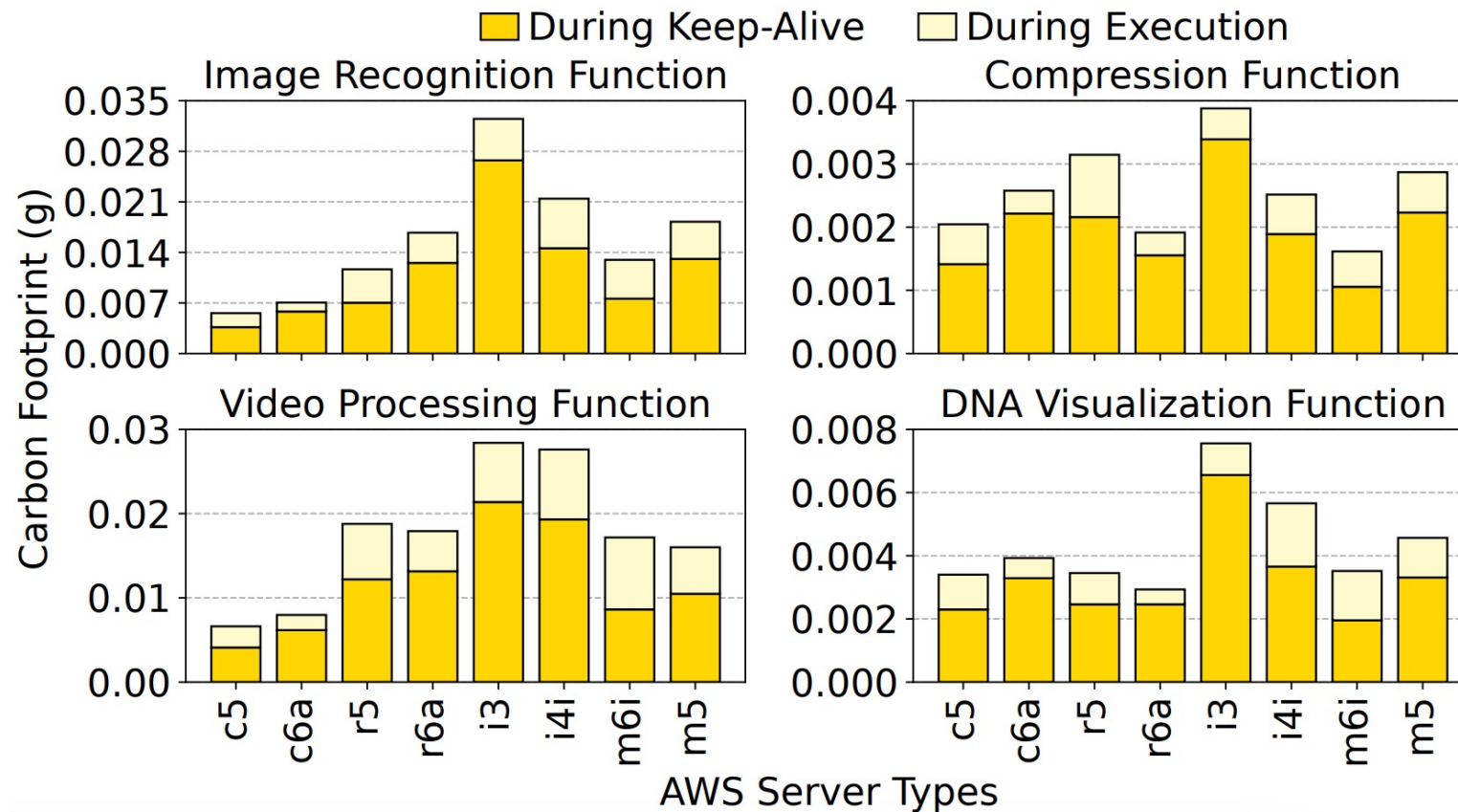
Choice B: DRAM embodied and operational for keep-alive carbon, divided among co-located functions

Choice C: DRAM embodied and operational for keep-alive carbon, CPU embodied for keep-alive

Choice D: DRAM embodied and operational for keep-alive carbon, CPU embodied and operational for keep-alive

Carbon estimation varies with methodology; ignoring DRAM embodied carbon and CPU keep-alive underestimates serverless carbon footprints

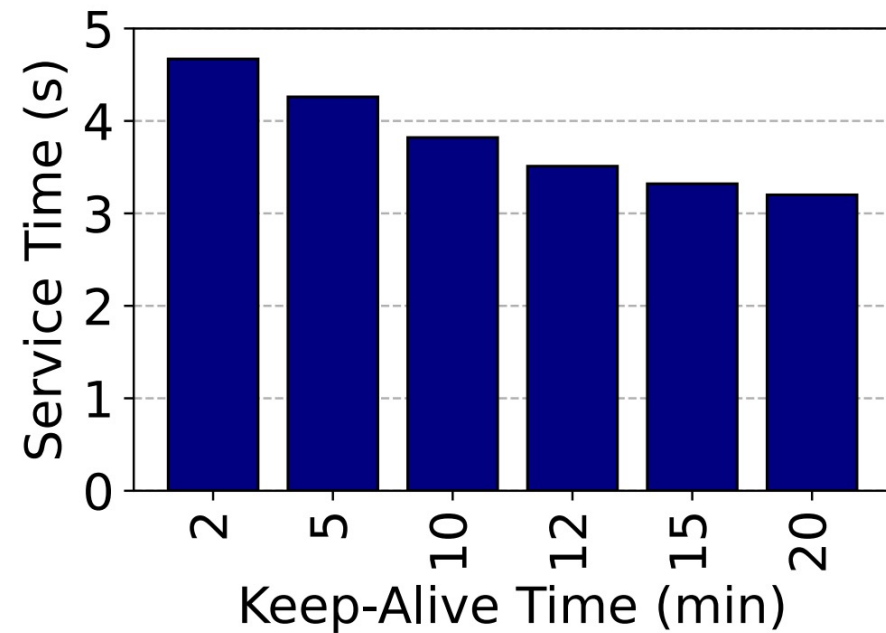
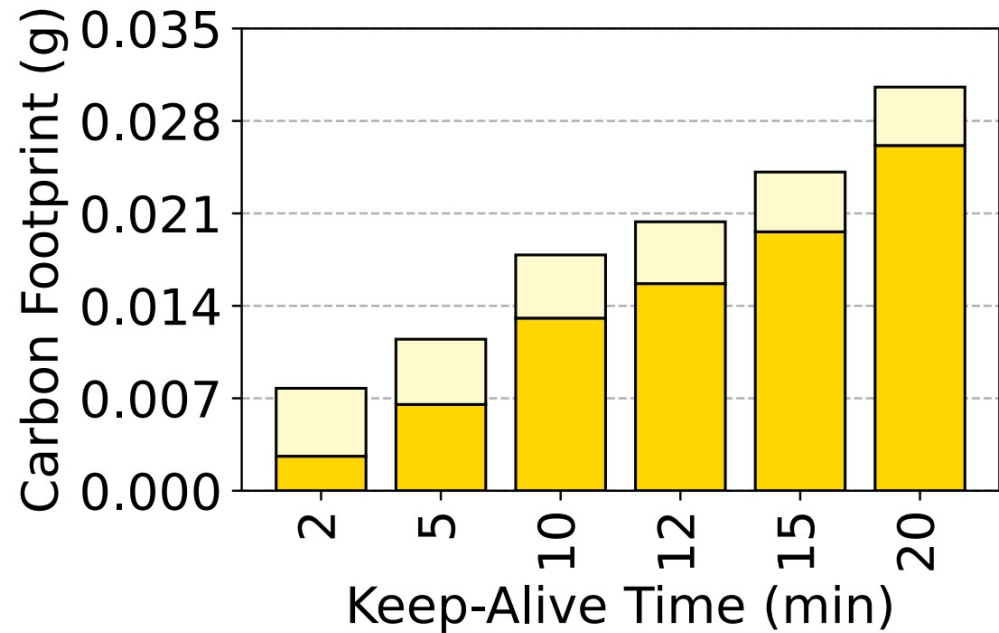
Scheduling in heterogeneous hardware can bring opportunity to reduce the carbon footprint



Carbon-optimal hardware varies by function and often differs from energy-optimal, shorter keep-alive durations suit newer hardware to reduce embodied carbon impact.

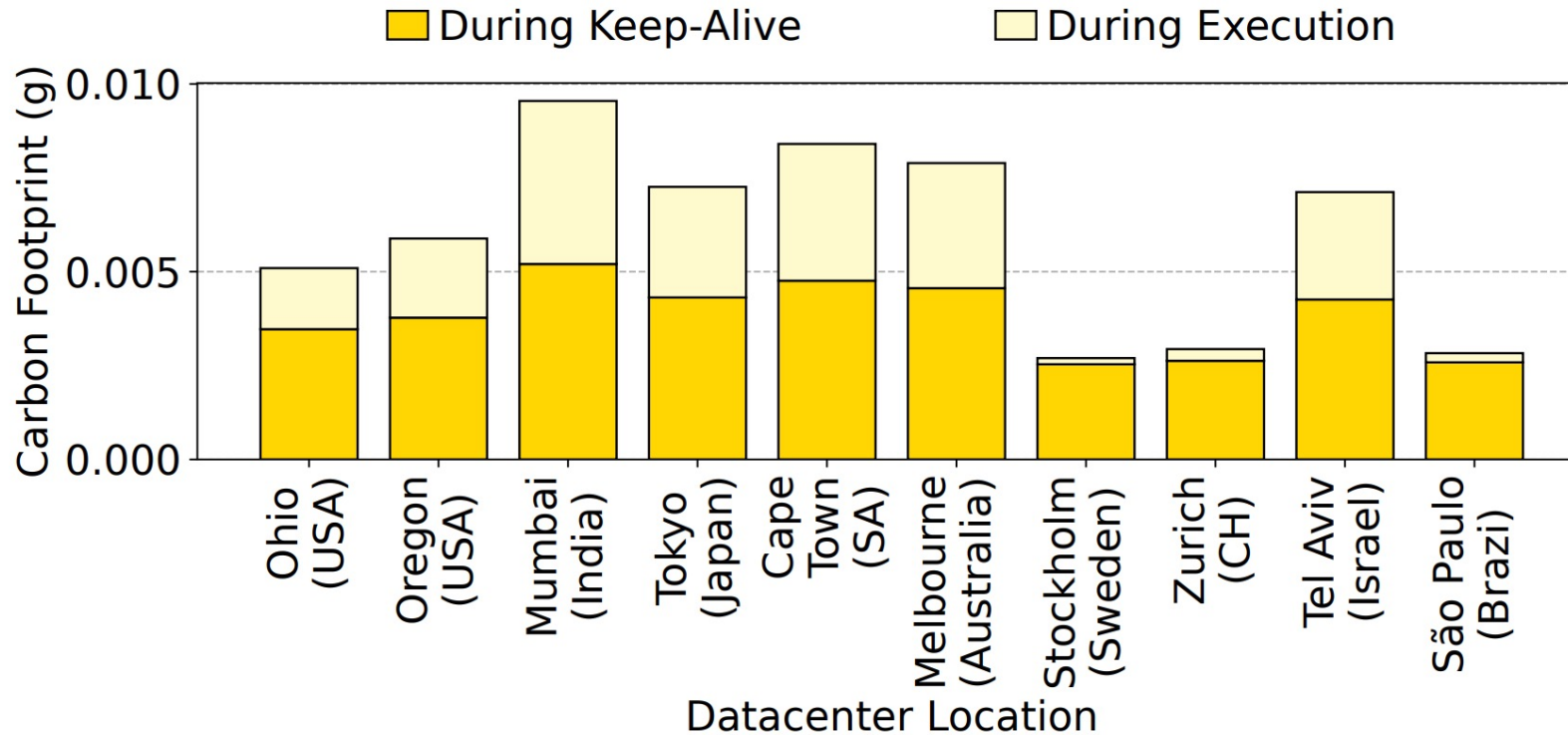
Performance and carbon footprint are at odds with respect to the keep-alive time of functions

■ During Keep-Alive ■ During Execution



Place high keep-alive functions on older hardware and high-energy functions on newer hardware

Carbon footprint varies across different datacenter locations due to variations in carbon intensity



Use older hardware with lower embodied carbon for keep-alive in low-carbon regions, and energy-efficient hardware for execution in high-carbon regions

Carbon footprint accounting methodologies should be standardized in serverless computing

The Hidden Carbon Footprint of Serverless Computing

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ABSTRACT

Due to the unique aspects of serverless computing like keep-alive and co-location of functions, it is challenging to account for its carbon footprint. This is the first work to introduce the need for systematic methodologies for carbon accounting in the serverless environment, propose new methodologies and in-depth analysis, and highlight how the carbon footprint estimation can vary based on the chosen methodology. It discusses how serverless-specific scheduling choices can impact the trade-offs between performance and carbon footprint, with an aim toward standardizing methodological choices and identifying opportunities for future improvements.

CCS CONCEPTS

• Computer systems organization → Cloud computing

KEYWORDS

Serverless Computing, Carbon Footprint, Sustainability

ACM Reference Format:

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1 Introduction And Background

Carbon emissions of computing systems. The greenhouse gas emissions from large-scale computing systems constitute more than 2.5% of global emissions [23, 29]. With the increasing demand for computing resources, the carbon footprint of computing is also increasing rapidly. Industries and cloud computing service providers are increasingly aiming to enhance the sustainability of their operations, focusing

on reducing carbon emissions during the manufacturing of computing systems, energy consumption during operations, and using renewable energy during operations.

The carbon footprint of computing systems comprises the operational and embodied carbon. The operational carbon is the carbon emitted during the hardware operation, for the energy required for operation. It is a product of the energy consumption and the carbon intensity (CI). The CI is the amount of carbon emitted per unit of generated energy from a specific source (i.e., if a power grid uses more renewable sources at a time, it delivers energy with lower CI). The CI varies temporally and spatially depending on the energy mix being used by a power grid of a location. The embodied carbon is the carbon emitted during the manufacturing of hardware, and is amortized over the lifetime of hardware. Due to the effect of embodied carbon during manufacturing and varying levels of carbon intensity, minimizing carbon footprint is not the same as minimizing energy consumption, as also acknowledged by prior studies [21, 23] – although minimizing energy consumption is certainly helpful in also reducing carbon footprint.

Serverless computing and its carbon footprint. Serverless computing is a rapidly increasing form of cloud computing, with more than 70% of users of production cloud platforms like AWS, using one or more serverless solutions [11]. Since serverless functions are typically short-running, providers keep alive functions in the memory of servers when they are predicted to be invoked. If a function is not kept alive and it is invoked, it undergoes a cold start, otherwise, it undergoes a warm start. A higher keep-alive period increases the warm start chances. The summation of the execution time and the cold start time (0, if warm start) is the service time of a function.

While there have been attempts to measure and model the carbon footprint of applications running on a dedicated environment, they cannot be directly translated to estimate the carbon footprint of serverless computing. In this work, we argue that serverless computing has considerable environmental implications. While the stateless nature of serverless computing is attractive, the need to keep alive functions in memory incurs high embodied carbon emissions and should be accurately and robustly accounted for. Furthermore, the

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Recommendations for providers to perform carbon aware scheduling

Methodologies to measure carbon in the cloud

Insights on serverless carbon variation with application and hardware types

Future research should consider optimizing for both performance and sustainability, ensuring cross-platform consistency

