

# Algorithmic Foundations of Grid-Integrated AI Infrastructure

*December 2025 @NSF Compute-Energy Nexus  
Paper presented at ACM SIGCOMM 2025*



Adam Lechowicz, Rohan Shenoy, Noman Bashir, Mohammad Hajiesmaili, Adam Wierman, Christina Delimitrou



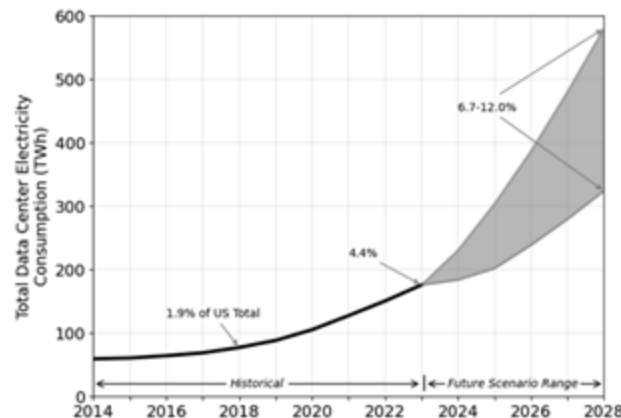
Caltech



Berkeley  
UNIVERSITY OF CALIFORNIA

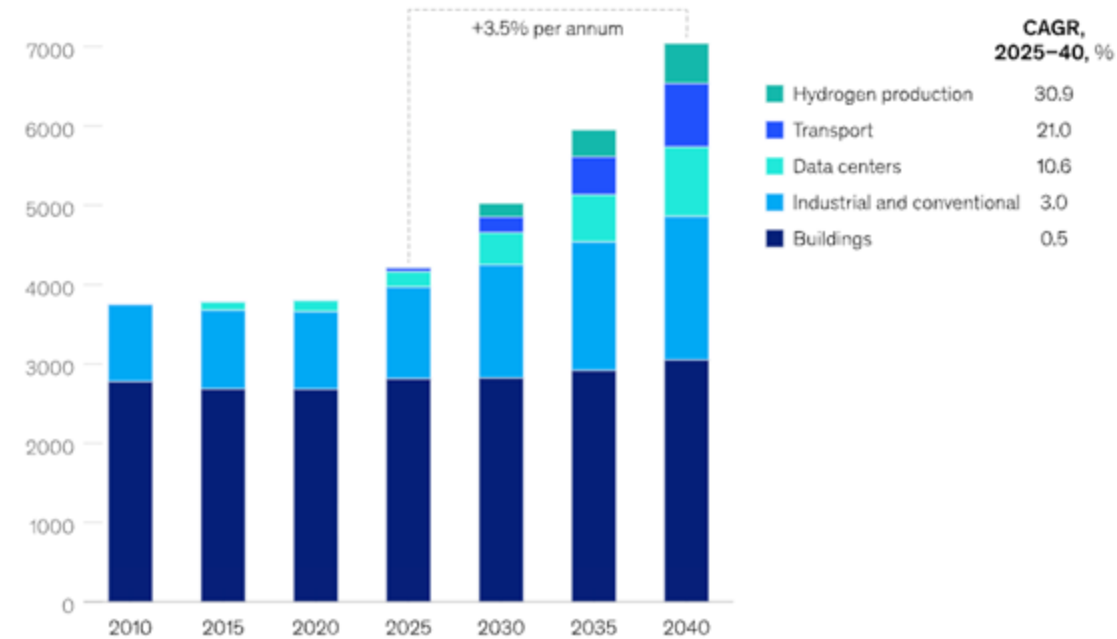
# US power demand is skyrocketing

- **Unprecedented** (absolute) demand growth
  - 150–250 TWh per year
- Data centers, running **AI workloads**, are expected to add ~450 TWh/yr by 2040, 10–12% of total growth.



US power demand is expected to increase up to 3.5 percent annually through 2040.

US power demand, by sector, current trajectory scenario,<sup>1</sup> terawatt-hours



<sup>1</sup>Volumetric power demand load is total downstream sales, comparable to data that operators are required to provide on the US Energy Information Administration's Form EIA-861. It represents the gross of distributed generation, behind-the-meter systems, and transmission and distribution losses. Source: McKinsey Power Model

# AI is forcing a redesign of the compute backbone

and calls for the integration of **compute** and **energy** infrastructure

- AI workloads are pushing compute, cooling, and power systems **beyond traditional limits**
- Power access is becoming a **primary** bottleneck for scaling AI compute
- **Integration** with energy infrastructure (renewables, microgrids, storage) is essential

Why the AI era is forcing a redesign of the entire compute backbone

Amin Vahdat, Google

August 3, 2025

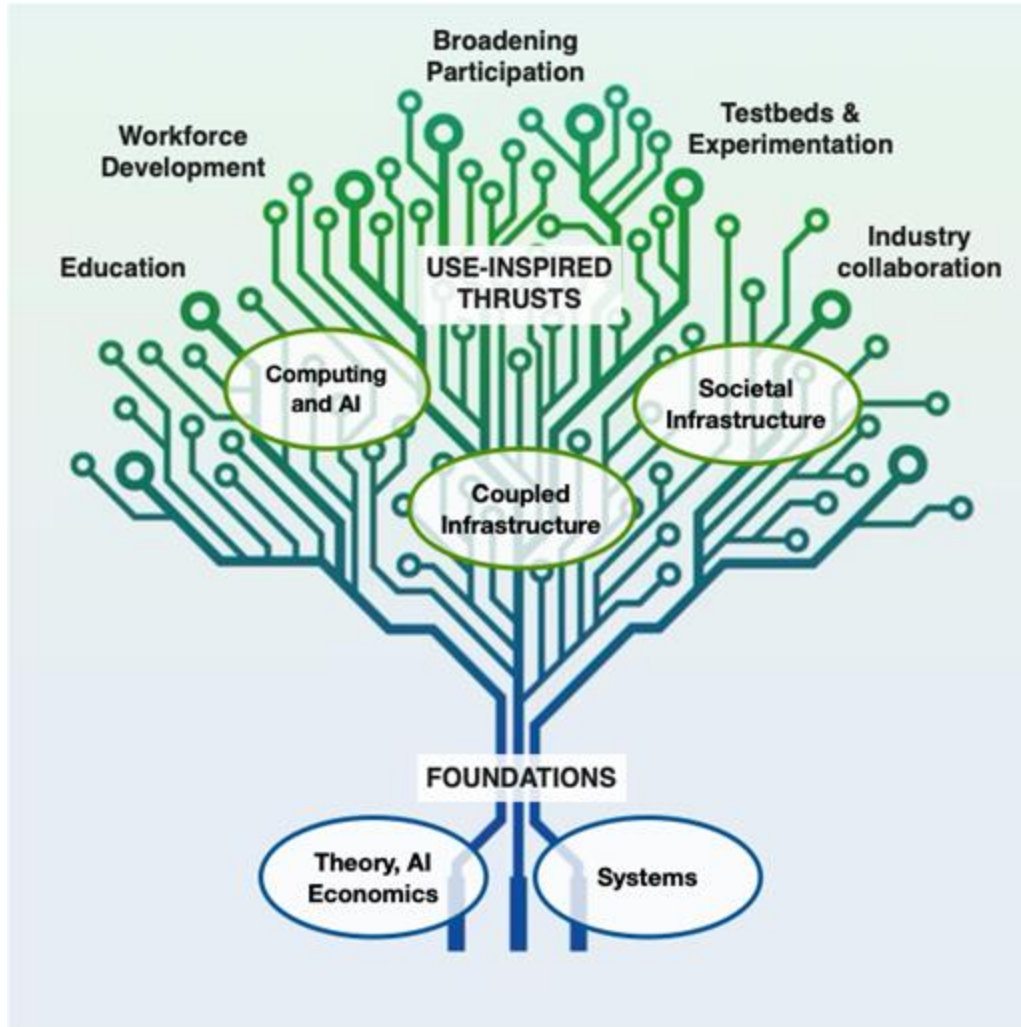


- To meet **AI's massive scaling**, we must scale energy supply, delivery, and **adaptiveness**
- e.g., **turning down compute** when power is constrained and using real-time telemetry
- otherwise, efficiency, cost, and feasibility will suffer

**concerns about the environmental impacts of this massive growth ...**

# CoDec: a new field of **computational decarbonization**

*to unify, transform, and accelerate societal decarbonization across domains*



## Theory and AI Foundations

- Learning & optimization, Incentive design

## Systems Foundations

- Carbon services, coordination abstractions

## Decarbonize Computing and AI

- AI, data center servers, clients, edge networks

## Decarbonize Societal Infrastructure

- Built environment, transportation, human factors

## Coupled Infrastructures

- Grid, computing couplings, buildings + transport

# Computational decarbonization of societal infrastructures

The CoDec Expedition will **develop the new field of Computational Decarbonization**, focusing on optimizing and reducing the lifecycle carbon emissions of complex computing and societal infrastructure systems.



**Sensing** to provide visibility into systems' operational and embodied carbon over their lifetime



Exploit new dimensions of **energy flexibility** in modern infrastructure for optimizing carbon-efficiency



Software-defined interfaces and systems for programmatically **deploying these optimizations** to reduce carbon emissions



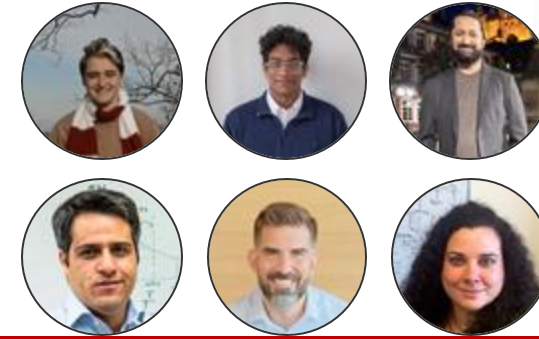
Intuitive interfaces that **account for concerns of privacy**, fairness, equity, incentives, and user preference

# Cost of flexibility!



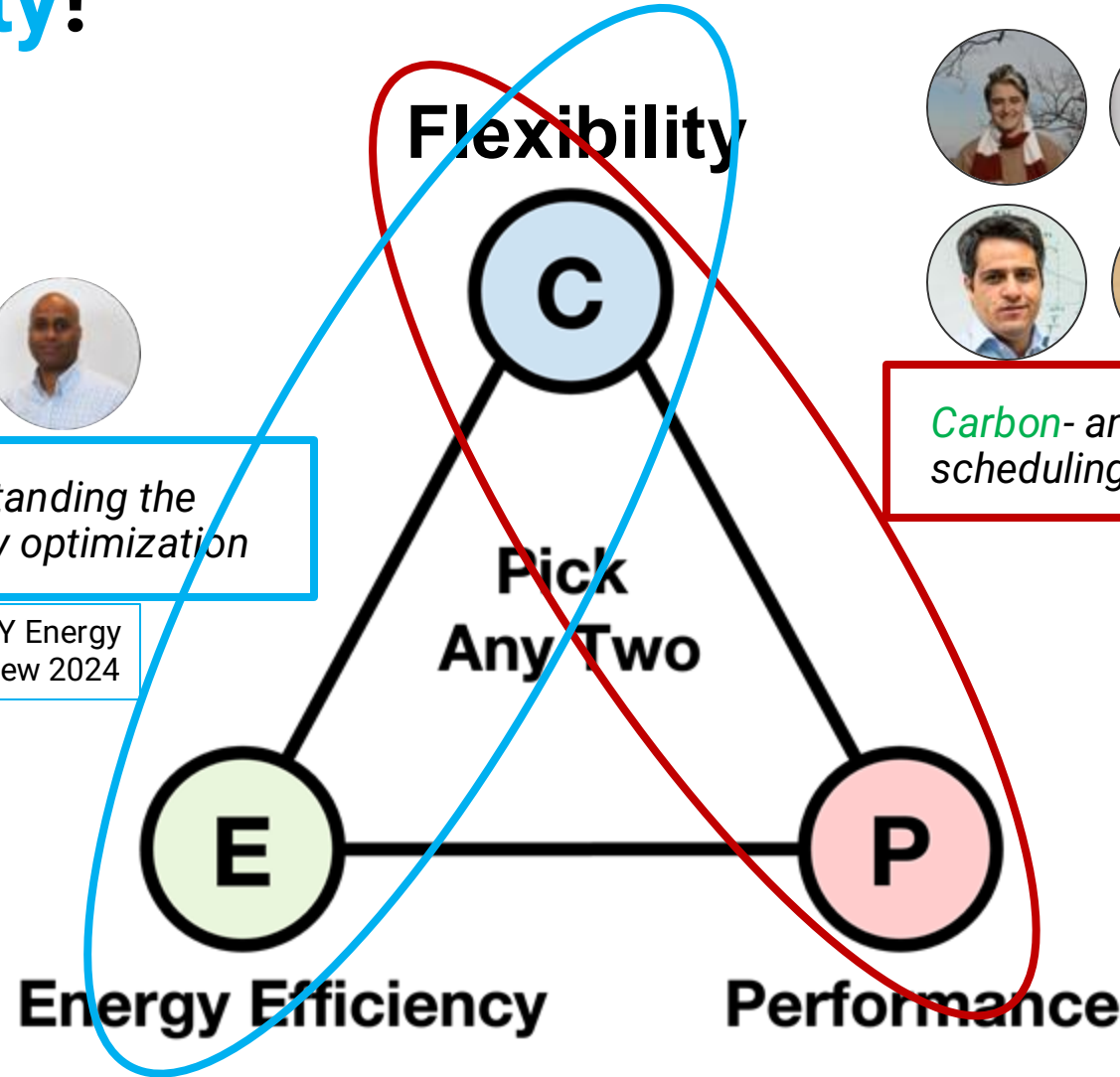
*The war of the efficiencies: Understanding the tension between carbon and energy optimization*

ACM SIGENERGY Energy Informatics Review 2024



*Carbon- and precedence-aware scheduling for data processing clusters*

SIGCOMM 2025

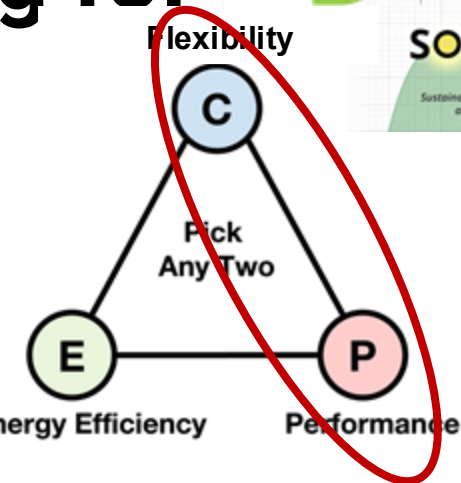


- W. Hanafy, et al. "The war of the efficiencies: Understanding the tension between carbon and energy optimization." in ACM SIGENERGY Energy Informatics Review 2024
- R. Bostandoost, et al. "Data-driven Algorithm Selection for Carbon-Aware Scheduling." in ACM HotCarbon 2024
- A. Lechowicz, et al. "Carbon- and Precedence-Aware Scheduling for Data Processing Clusters." in ACM SIGCOMM 2025 – short paper – full paper under review

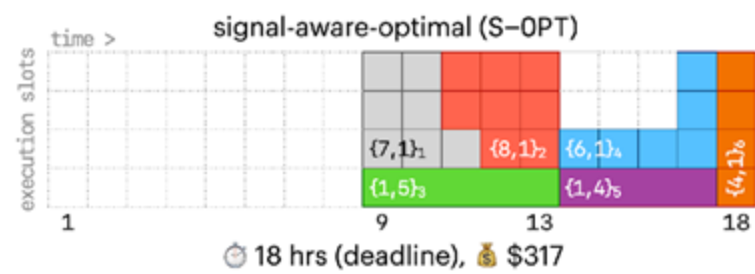
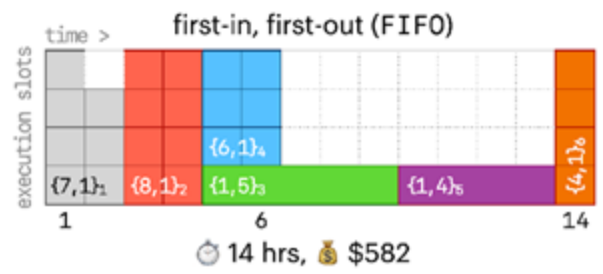
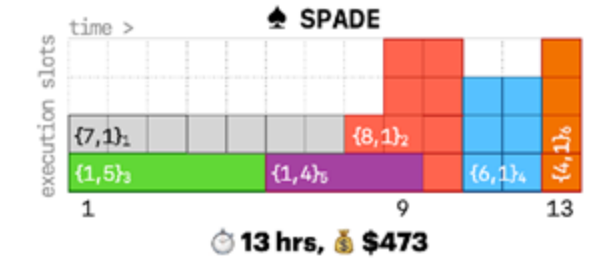
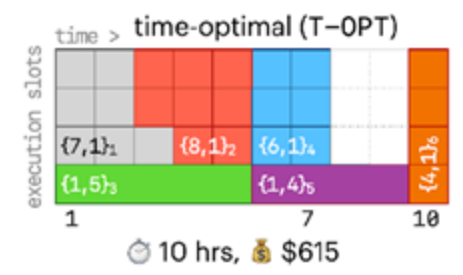
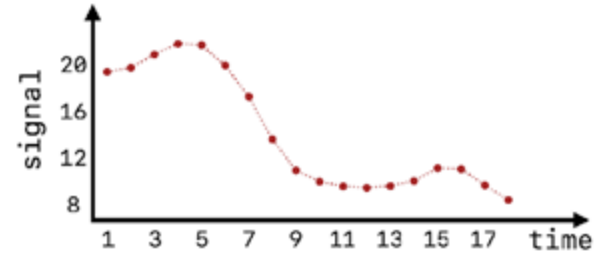
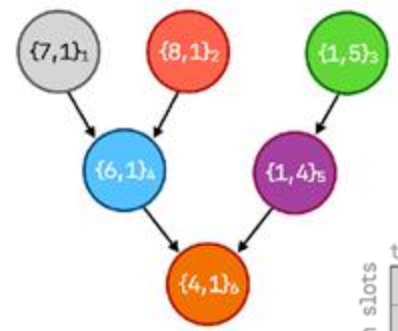
# Signal- and precedence-aware scheduling for data processing clusters



33% of the **lifetime carbon footprint** of a foundation model can be attributed to *data processing workloads*.



{# tasks, duration}<sub>ID</sub>



# Our main system: **SPADE**

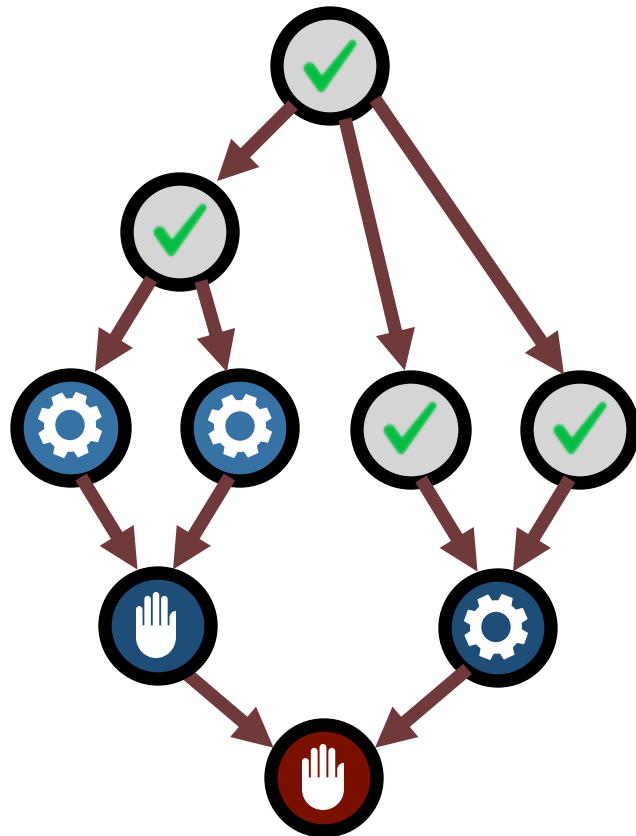
*Scheduling and Provisioning for Adaptive DAG Execution*



- **Key idea for joint scheduling and provisioning:** introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal

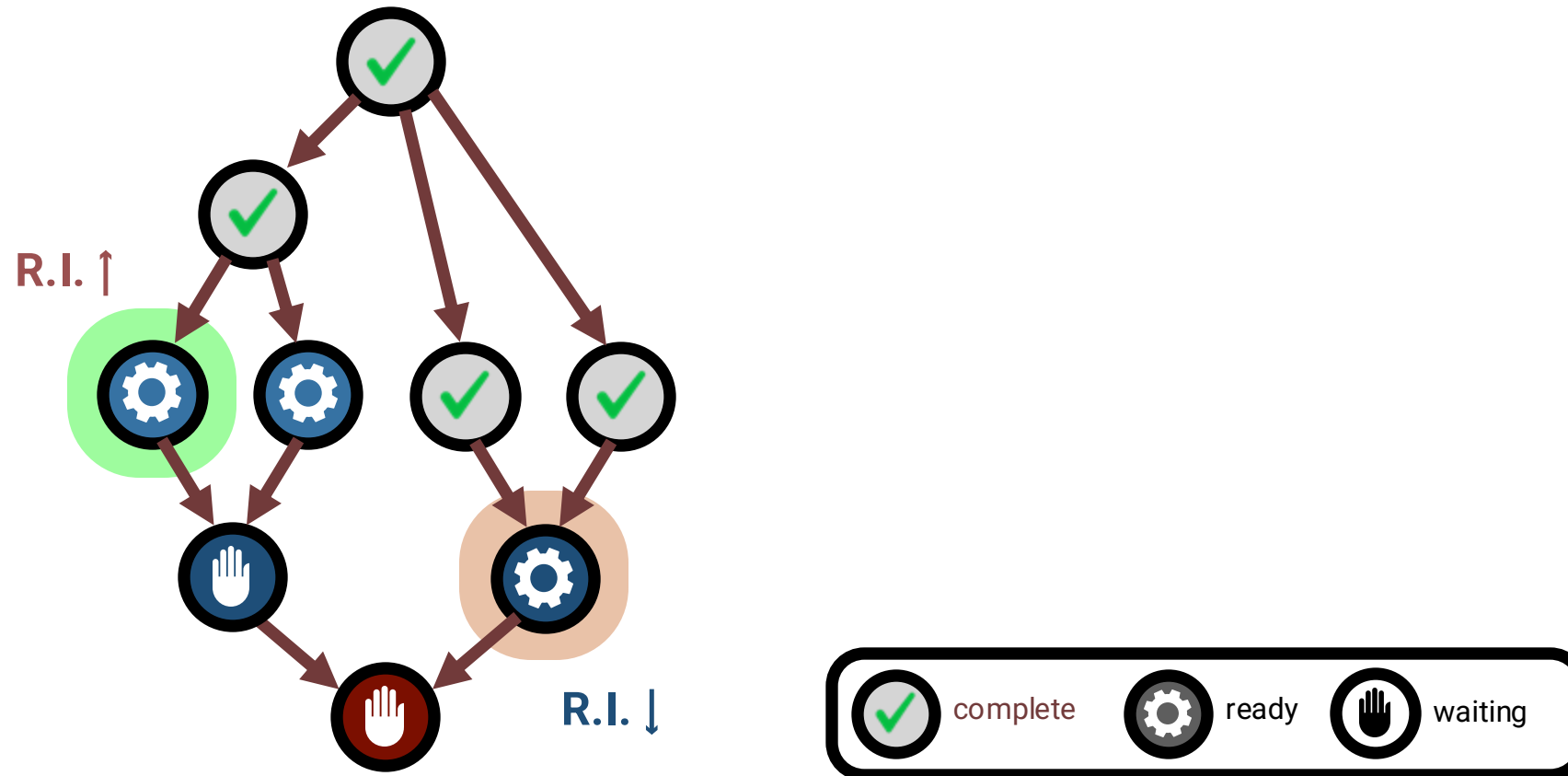
# Our main system: **SPADE**

- Key idea for joint scheduling and provisioning: introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal



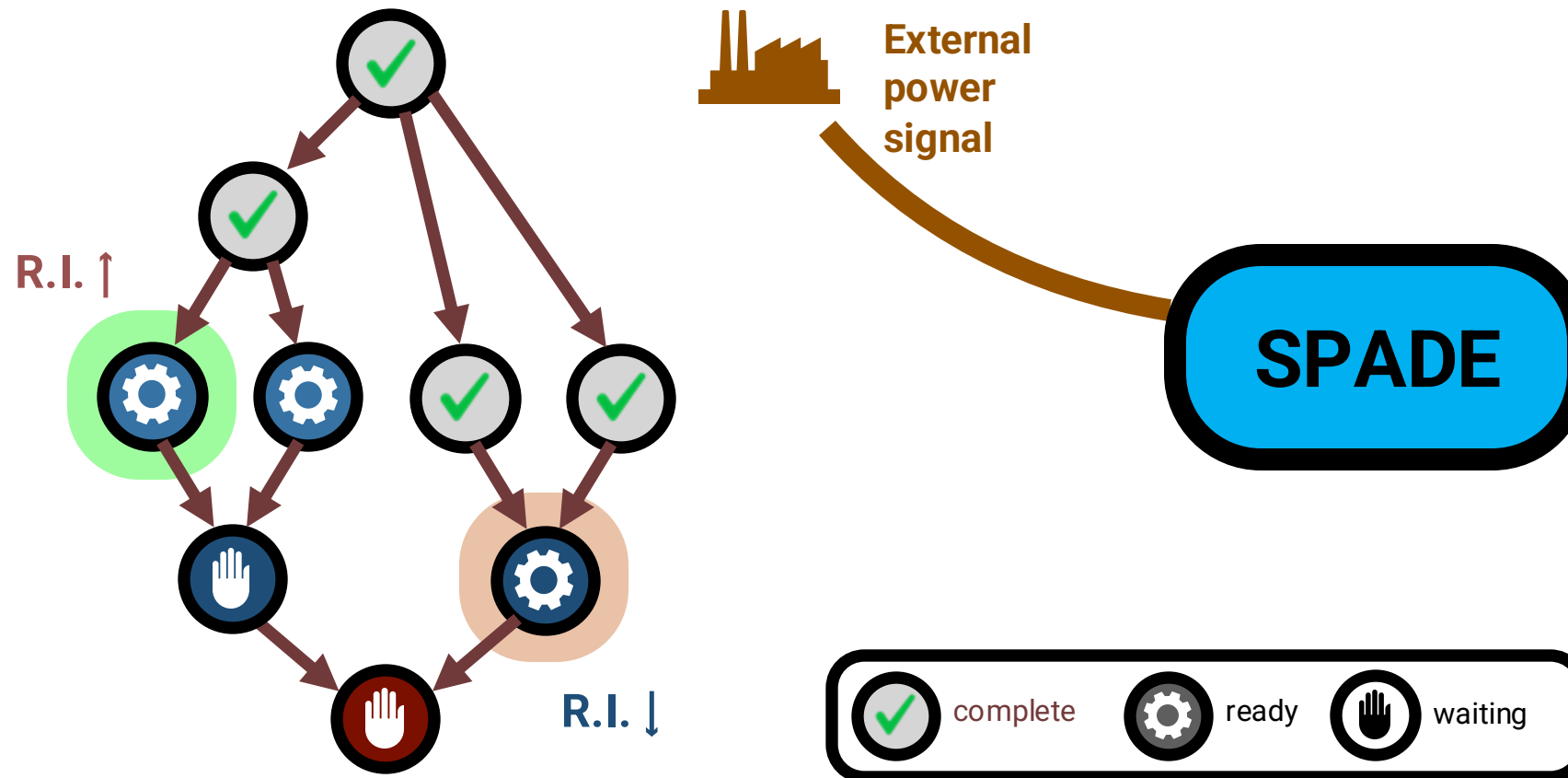
# Our main system: **SPADE**

- Key idea for joint scheduling and provisioning: introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal



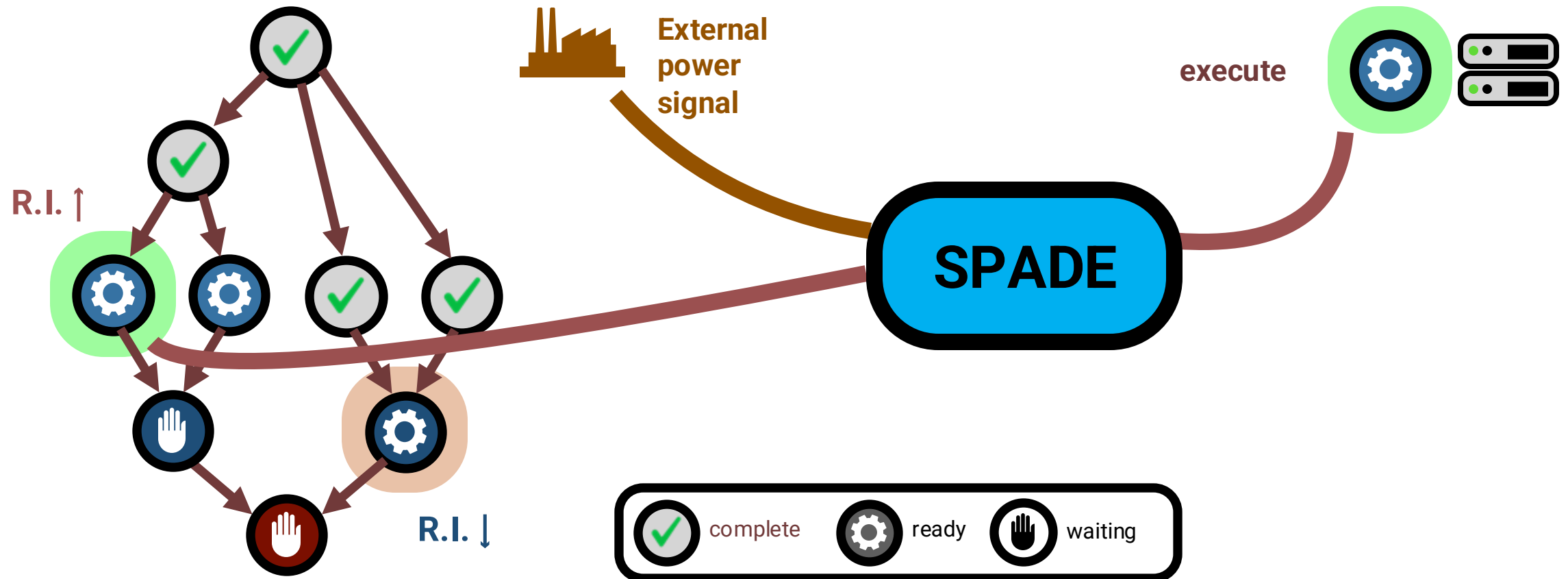
# Our main system: **SPADE**

- Key idea for joint scheduling and provisioning: introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal



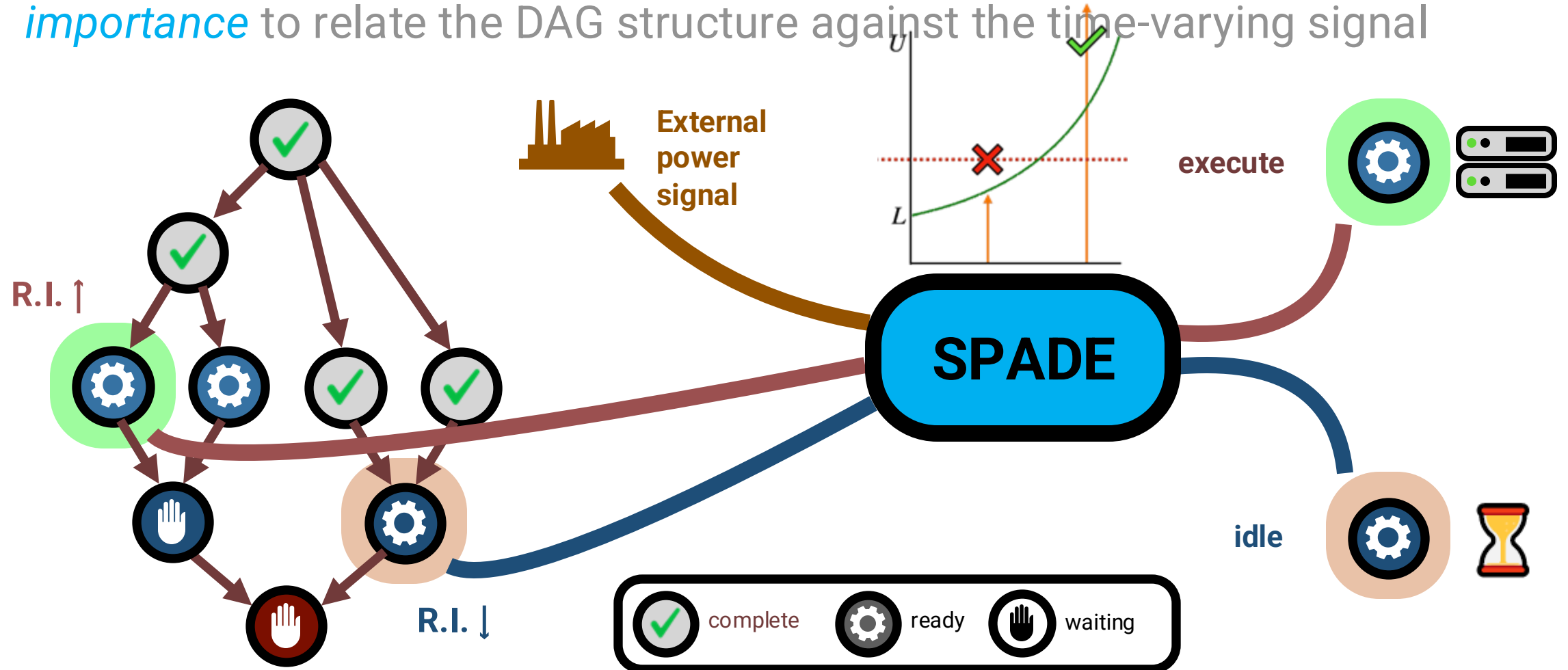
# Our main system: SPADE

- Key idea for joint scheduling and provisioning: introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal



# Our main system: SPADE

- Key idea for joint scheduling and provisioning: introduce a metric of *relative importance* to relate the DAG structure against the time-varying signal

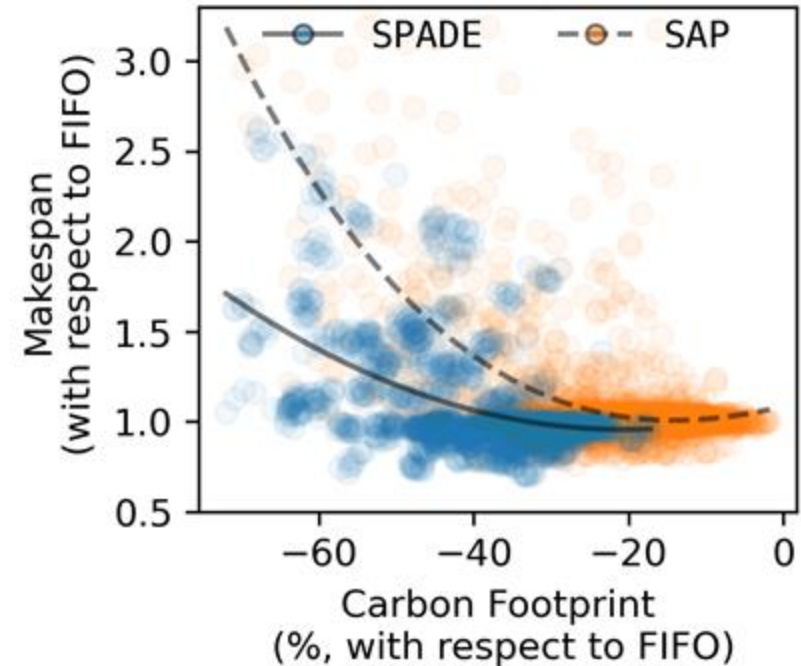


# Theoretical results

- ⌚ **Stretch Factor:** for a signal-aware schedule, the worst-case (multiplicative) increase in makespan compared to a signal-agnostic schedule.
- ⌚ [Informal Theorem]: Given a time-varying signal  $s$  and  $K$  resources, the stretch factor of **SPADE** is  $\leq 1 + \frac{D(\gamma, s)K}{2^{-1/K}}$  for a *signal-awareness parameter*  $\gamma \in [0, 1]$ 
  - ⌚  $D(\gamma, s)$  describes the *fraction of tasks shifted because of a high signal*
  - ⌚ As  $\gamma$  increases, the signal cost of **SPADE** decreases, at the expense of a larger stretch factor in the worst case.

# Performance evaluation

- Implemented **SPADE** in simulation env. and 100-node k8s testbed
- **SPADE** reduces **carbon** footprint by 32.9% with <5% increase in overall makespan; more noticeable impact on tail JCT
- Strictly better trade-off between performance and carbon compared to "provisioning-only" approach (**SAP**)



# Flexible scheduling and resource allocation

- **The Online pause and resume problem**
  - Carbon-aware workload scheduling (Sigmetrics 2024)
  - Carbon-aware electric vehicle charging management (Sigmetrics 2024)
  - Carbon-aware spatiotemporal workload management (Sigmetrics 2025)



Photo from Google

- A. Lechowicz, et al. "The online pause and resume problem: optimal algorithms and an application to carbon-aware load shifting." In *ACM SIGMETRICS 2024*.
- A. Lechowicz, et al. "Online conversion with switching costs: Robust and learning-augmented algorithms." In *ACM SIGMETRICS 2024*.
- A. Lechowicz, et al. "Learning-augmented competitive algorithms for spatiotemporal online allocation with deadline constraints." In *ACM SIGMETRICS 2025*.

# Flexible scheduling and resource allocation

Theoretical Foundations (bridging two lines of research)

Convex body chasing  
*(local and short-term switching costs)*

Online packing/covering  
*(enforcing long-term capacity/budget or demand constraints)*

Foundational CoDec problems feature both requirements in a unified problem

# Conclusion



Yes, we are dealing with order-of-magnitude larger problems!



Yes, the problems are at the intersection of multiple infrastructures!



But, the problems calls for fundamentally different ideas and techniques to solve!

This talk: joint scheduling and provisioning for joint minimization of first-order metrics (job completion time), and second-order metrics (the cost of the external signal)



# Expeditions: Computational Decarbonization of Societal Infrastructures at Mesoscales (CoDec)

