

# Exploiting Temporal Diversity of Water Efficiency to Make Data Center Less "Thirsty"

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# A massive data center



Google's data center in Mayes County, Oklahoma

# Something we know

- Data centers are energy hogs
  - Combined energy usage of all data centers would rank 5<sup>th</sup> in the world if data centers were a country
  - Tens of millions of **\$\$\$** in annual operational costs
  - Responsible for 2% electricity usage and 1% greenhouse gas emissions
  - .....

# Something we may **not** know

- Data centers are very “**thirsty**”
  - 2013 Uptime Institute survey shows that a majority of large data centers use cooling towers to remove heat from server rooms to the environment
  - AT&T’s data center facilities consume **1 billion** gallons of water in 2012
    - Taking up **30%** of the entire company’s water consumption

# Google's cooling towers



Cooling towers in Google's Belgium data center



# Google is “drinking” water



Water evaporation in Google’s Oregon data center

# Something we may **not** know

- Data centers also consume water “**remotely**”
  - Electricity production consumes water
    - The **#1** water withdrawal among all sectors
    - U.S. national average **1.8 L/kWh**, excluding hydropower
  - Data centers account for approx. **1%** water withdrawal in the U.S.
    - Data centers consume 1.7—2.3% electricity
    - Electricity accounts for 40+% water withdrawal

# Electricity production consumes water

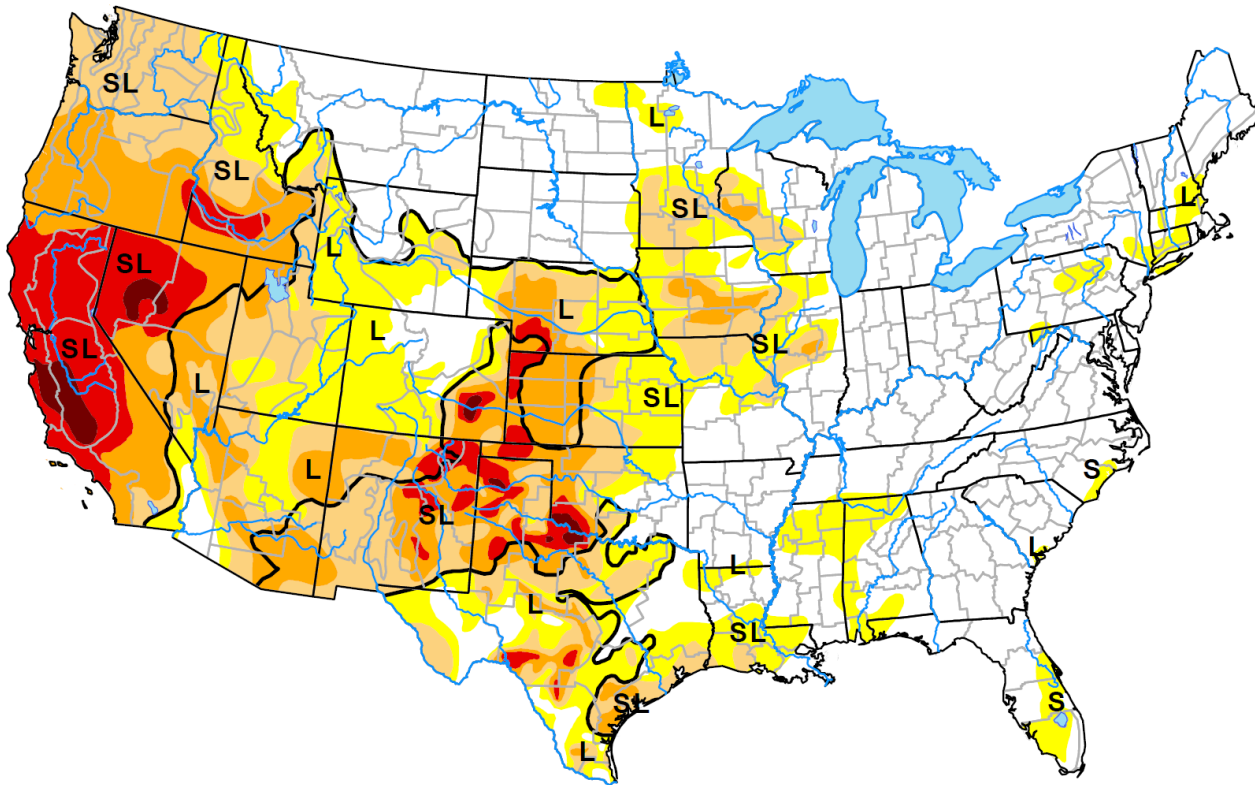


<http://www.globalenvision.org/>



# Why should we care about water?

- Extended droughts are becoming a norm



<http://droughtmonitor.unl.edu/>

# Why should we care about water?

- Extended droughts are becoming a norm
  - California declared state-wide drought emergency on Jan. 17, 2014, urging all residents and businesses to cut back water usage by **20%**

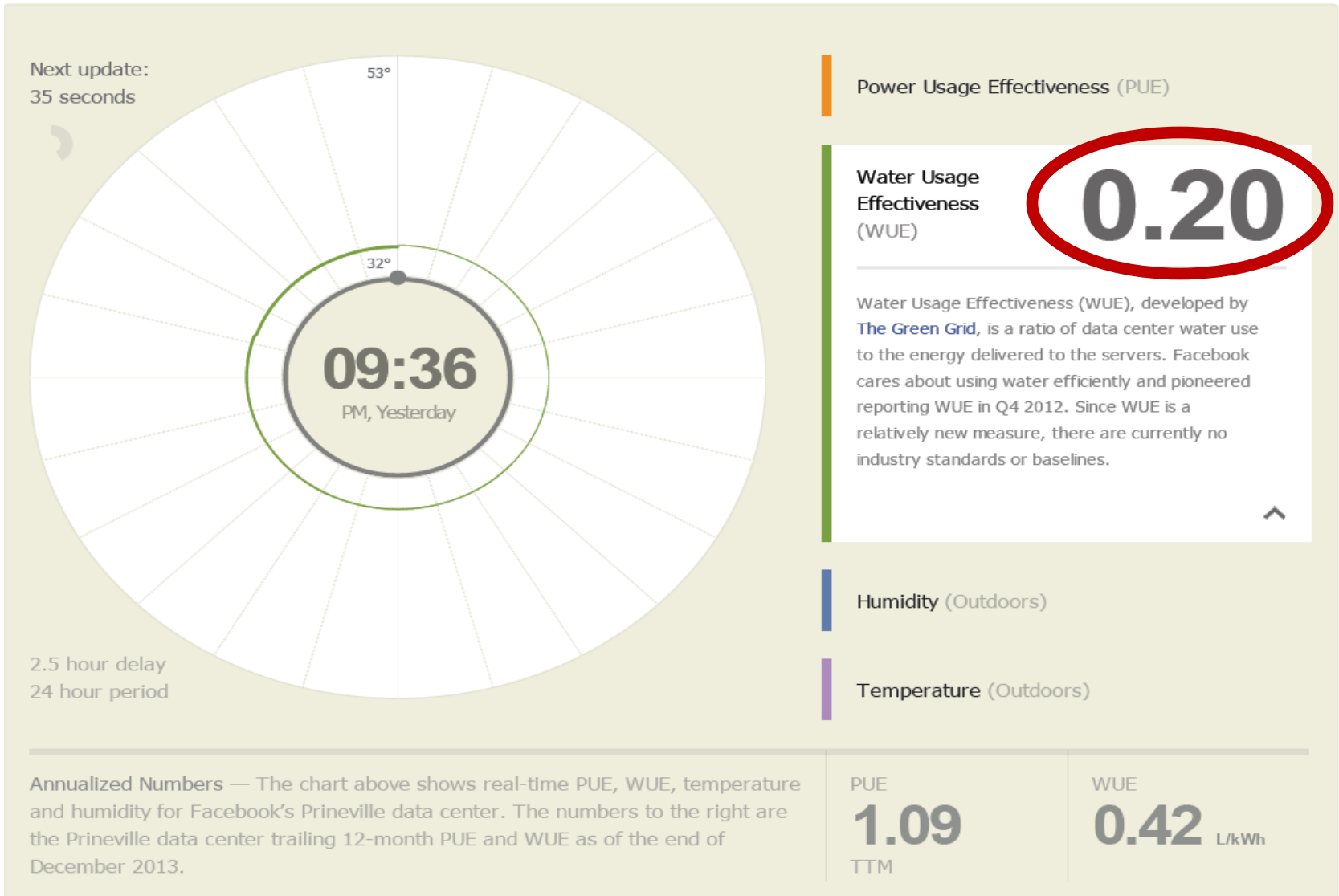
# What if drought is far away?

- Water conservation gets you...
  - Green certifications (e.g., LEED program)
    - Water reduction is a prerequisite
  - Tightening water compliance codes
    - U.S. government mandates 2% water efficiency improvement for federal facilities each year through 2020
  - Corporate social responsibility
    - AT&T, Google, Facebook, etc.

# Industry has realized the water issue

- Google uses recycled water/seawater in some of its facilities
- Facebook uses air economizer (a.k.a. “free” cooling)
  - Recently released source code for monitoring water usage
- In 2013, LBNL recommends **water** and **energy** as two key considerations for U.S. federal data center consolidation

# Industry has realized the water issue



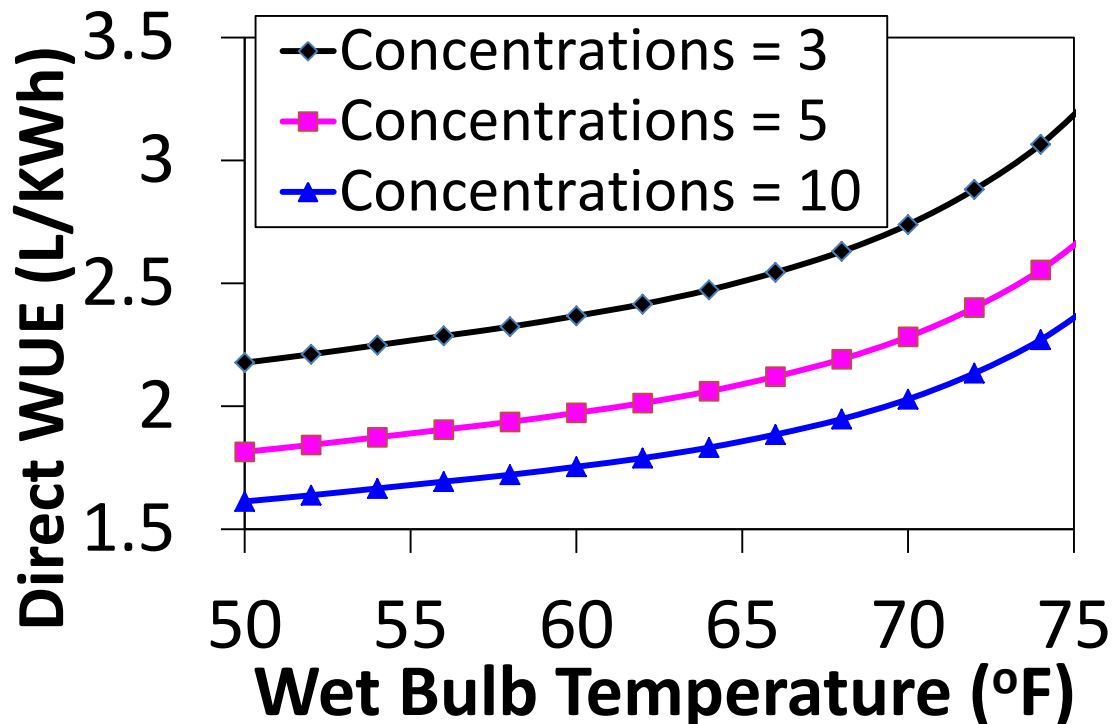
# Limitations of existing approaches

- The existing approaches primarily focus on improved “**engineering**”
  - High upfront capital costs
  - Suitable climate/locations
  - Offsite water is not explicitly addressed



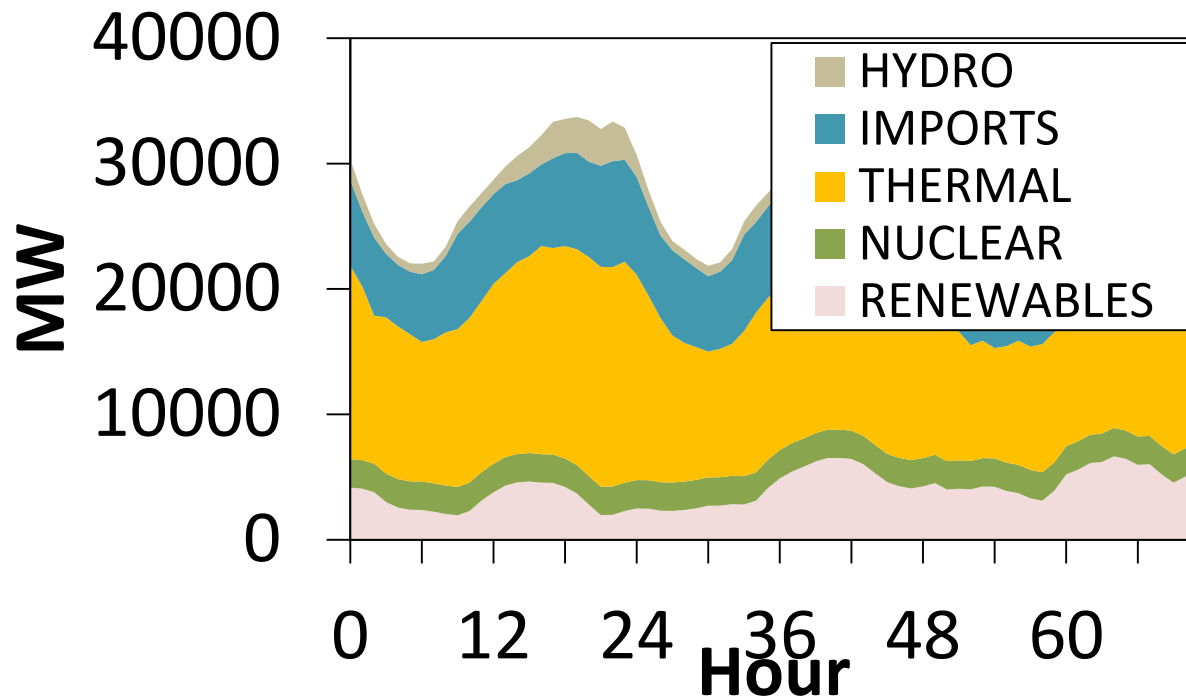
# Water is equal to energy?

- Not really...
  - Water efficiency changes over time
  - Just like minimizing electricity cost with time-varying electricity prices, we need to consider **“when”** to consume energy



# Time-varying (**offsite**) water efficiency

- Energy fuel mix changes over time for electricity
  - Different energy fuel types have different water efficiencies
  - In general, Hydro > Nuclear > Thermal > Renewables



**How can we exploit **time diversity** of water efficiency?**

# We propose...

- We can turn off some servers and defer some delay-tolerant workloads to time periods that are more water-efficient
  - But how to determine if a time period is **efficient**?
  - We also need to consider electricity cost (and carbon footprint if applicable)
- **WACE**: opportunistically schedule batch jobs
  - First, queue up the job arrivals
  - If the queue length is too large and/or water efficiency is already good, then process waiting jobs

# How does **WACE** work?

## – Input:

- Online information: WUE, electricity price, etc.

## – Solve online optimization:

- Minimize  $V \cdot g(t) - q(t) \cdot b(t)$ 
  - $g(t)$ : weighted sum of water footprint, electricity cost and carbon footprint
  - $q(t)$ : batch job queue length
  - $b(t)$ : processed batch jobs

## – Update batch job queue length

**Provably-efficient** while bounding queue length

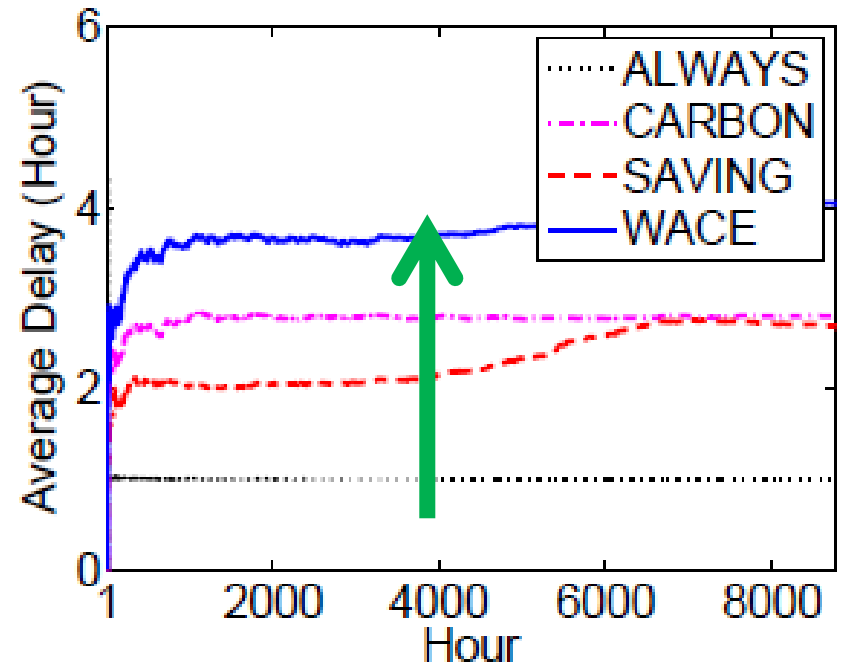
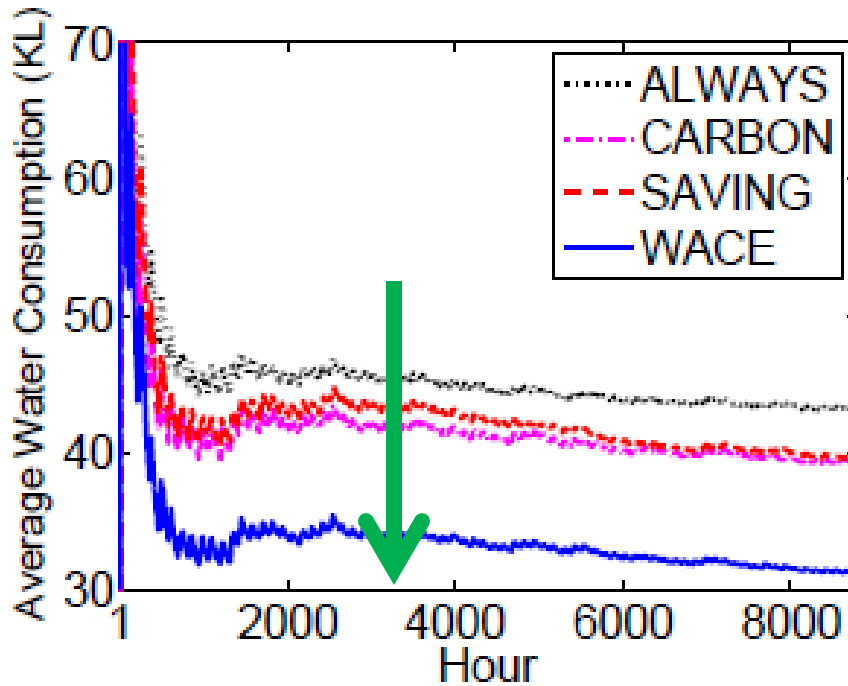
# Case study



# How to evaluate **WACE**?

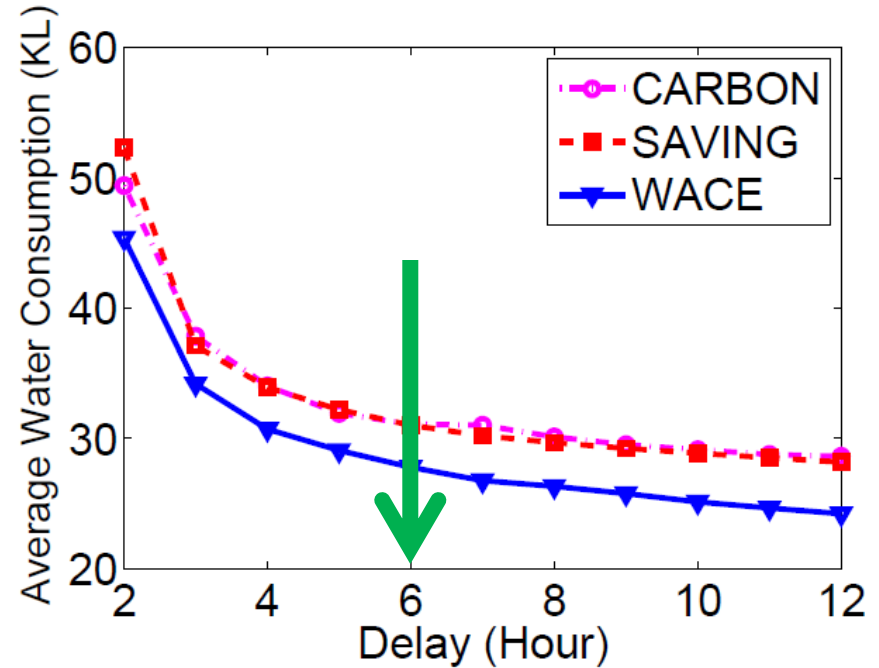
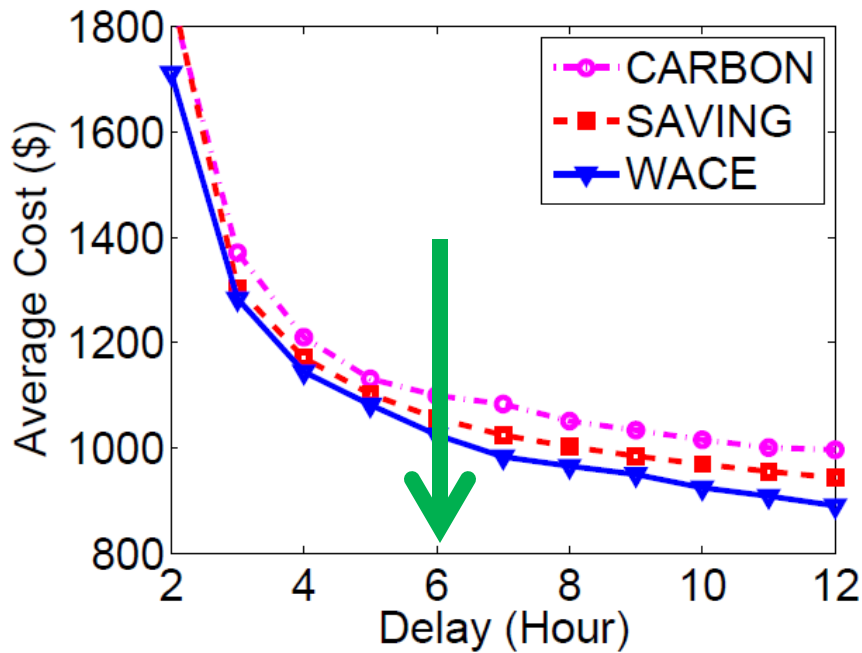
- Let's consider a real-life trace simulation
  - MSR workload; data center in Mountain View, California
- Benchmarks
  - **SAVING**: Optimize for electricity cost
  - **CARBON**: Optimize for carbon footprint
  - **ALWAYS**: Optimize for delay performance

# WACE reduces water footprint



- Big water saving at (a small) delay increase

# WACE reduces total cost



- WACE reduces the total cost without increasing delay
- Minimizing electricity cost or carbon emission is **not** sufficient for minimizing water footprint!

# Three messages

- Data centers are “thirsty”
- Water footprint **!=** energy or electricity cost or carbon emission
- Time diversity of water efficiency can be exploited for reducing water footprint



# Thanks!