

Will EVs Break the Power Grid?

Professor Scott Moura

PATH Faculty Director | Chair of Engineering Science

Clare and Hsieh Wen Shen Endowed Distinguished Professor

University of California, Berkeley

9 May 2025





Two More Chinese Companies Announce 'Megawatt' EV Charging

The new 1.2-megawatt and 1.5-megawatt chargers from Zeekr and Huawei are absolutely nuts. But are they really needed?





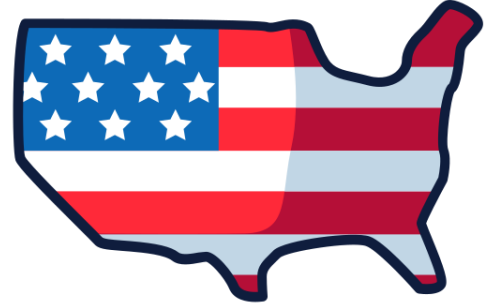
1
kilowatt



1
megawatt



1
gigawatt



5000
gigawatts

Charge
EV overnight



Charge EV
overnight, at home



Charge in
30 minutes

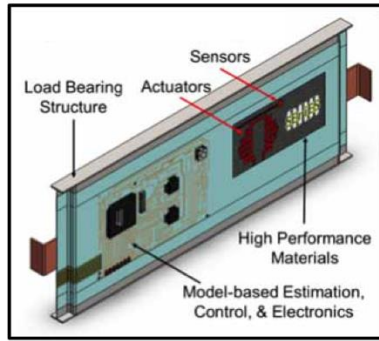


Charge as
quickly as filling gas



Charge faster
than a sneeze

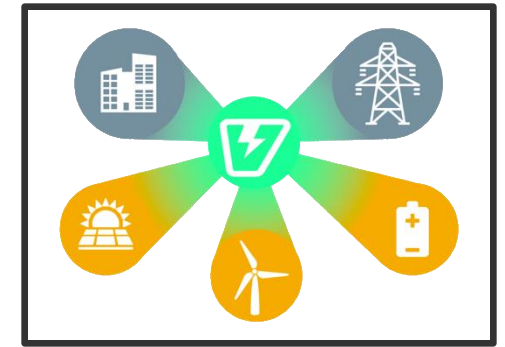




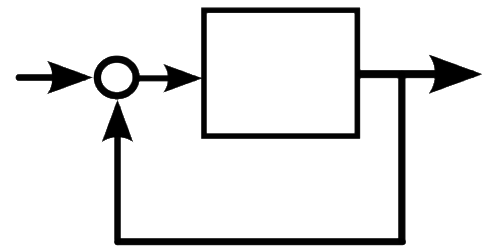
Battery
Management
Systems (#BATT)



Automated, Connected,
& Electric Vehicles
(#ACES)

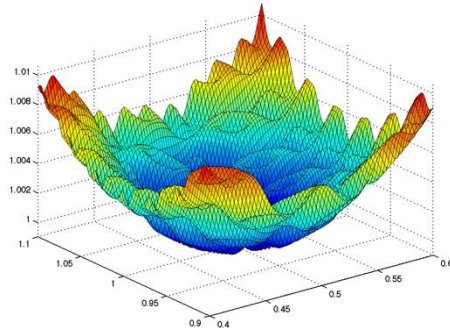


Power Systems,
Grids, Markets
(#GRID)



Dynamic Systems
& Control

Optimization



Data Science



eCAL Research Areas



Patrick
KEYANTUO



Tugba
OZTURK



Preet
GILL



Ruiting
WANG



Junzhe
SHI



Yi
JU



Shida
JIANG



Joyce
CHEN



Linda
LIM



Lejun
ZHOU



Jaewoong
LEE



Shengyu
TAO



Bhumi
TANDEL



Sam
MURTHY



Sam
BOBICK



Michael
HENDERSON



Jingchun
WANG



Zihe
LIU



Shannon
HOANG

OUR PRODUCT
IS PEOPLE

Alumni – PhD and Postdocs



Dr. Eric
BURGER
Nest / Google



Dr. Hector
PEREZ
John Deere



Dr. Caroline LE
FLOCH
Storio Energy



Dr. Eric
MUNSING
Tesla



Dr. Bertrand
TRAVACCA
GridBeyond



Dr. Laurel
DUNN
Morphosis



Dr. Sangjae BAE
Honda
Research
Institute



Dr. Saehong PARK
Apple



Dr. Mathilde
BADOUAL
Storio Energy



Dr. Zach GIMA
Form Energy



Dr. Dong
ZHANG
Asst Prof. at OU



Dr. Soomin
WOO
Asst Prof.
at Konkuk U



Dr. Teng ZENG
Tesla



Dr. Zhijia
HUANG
BYD



Dr. Hassan
OBEID
Attentive



Dr. Ioanna
KAWADA
PG&E



Dr. Dimitris
VLACHOGIANNIS
Uber



Dr. Aaron
KANDEL
Nissan USA



Dr. Dylan KATO
Form Energy



Dr. Chitra
DANGWAL
Archer Aviation



Dr. Guillaume
GOUJARD
Tesla



Prof. Azad
GHAFFARI
Wayne State
University



Prof. Xiaosong
HU
Chongqing
University



Prof. Satadru
DEY
Penn State
University



Prof. Chao SUN
Beijing Institute
of Technology



Prof. Hongcai
ZHANG
University of
Macau



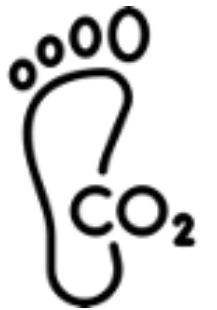
Dr. Milad
MEMARZADEH
NASA Ames

California Partners for Advanced Transportation Technology (PATH)



Mission:

Realize a **safe, equitable, efficient, and carbon-neutral transportation system** for all, through research and development of advanced ideas and technologies.

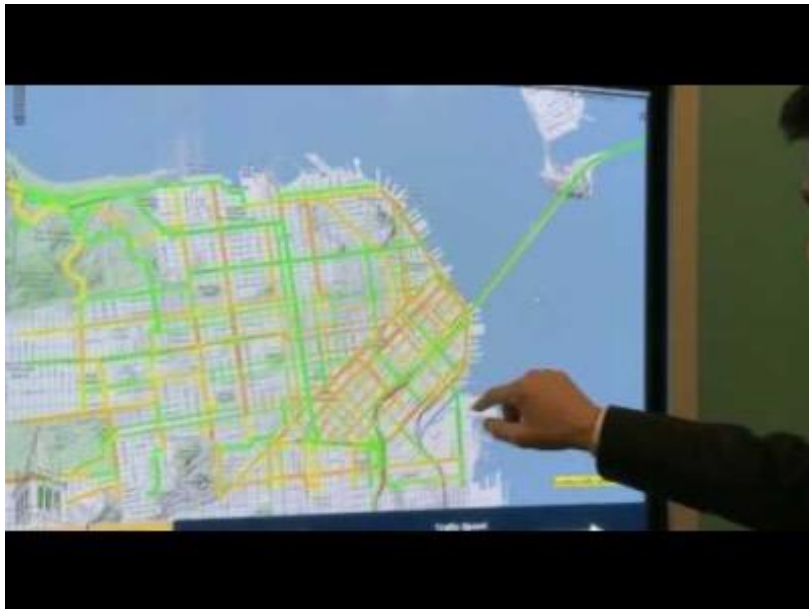


PATH: History of Transportation Innovation



←
1997 Automated car platooning
on I-15 near San Diego

→
2001 Performance Measurement
System (PeMS) collects real-time
data from ~40,000 road sensors



←
2009 Mobile Millennium tracks real-
time traffic via Nokia cell phones.
GMaps traffic b4 Google.

→
2016 Berkeley Deep Drive:
Industrial/ academic research consortium on
deep automotive perception technologies



PATH's Role in Transportation Technology Ecosystem

Engine of Innovation: Research → Development → Demonstration → Deployment

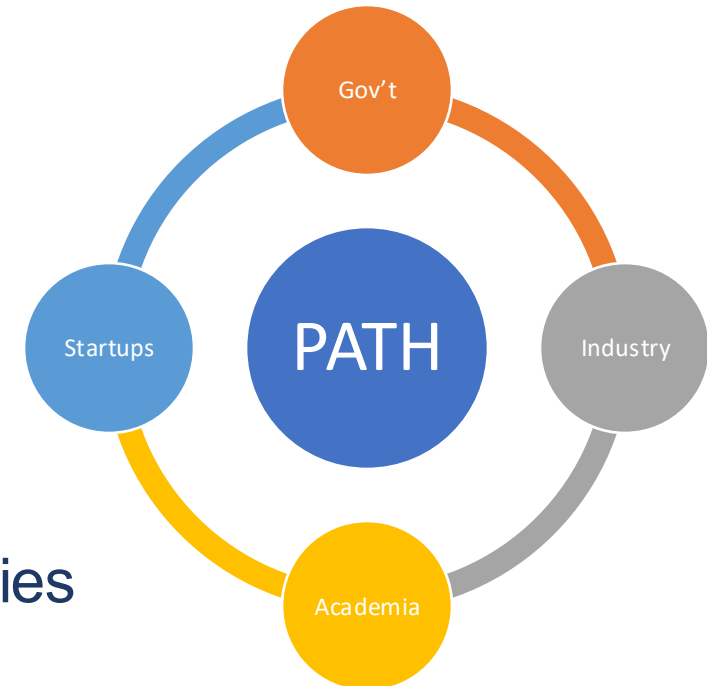


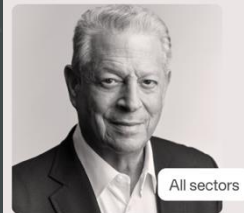
Berkeley
UNIVERSITY OF CALIFORNIA



Start-ups, Big Industry, Government Agencies, etc.

Build
Communities

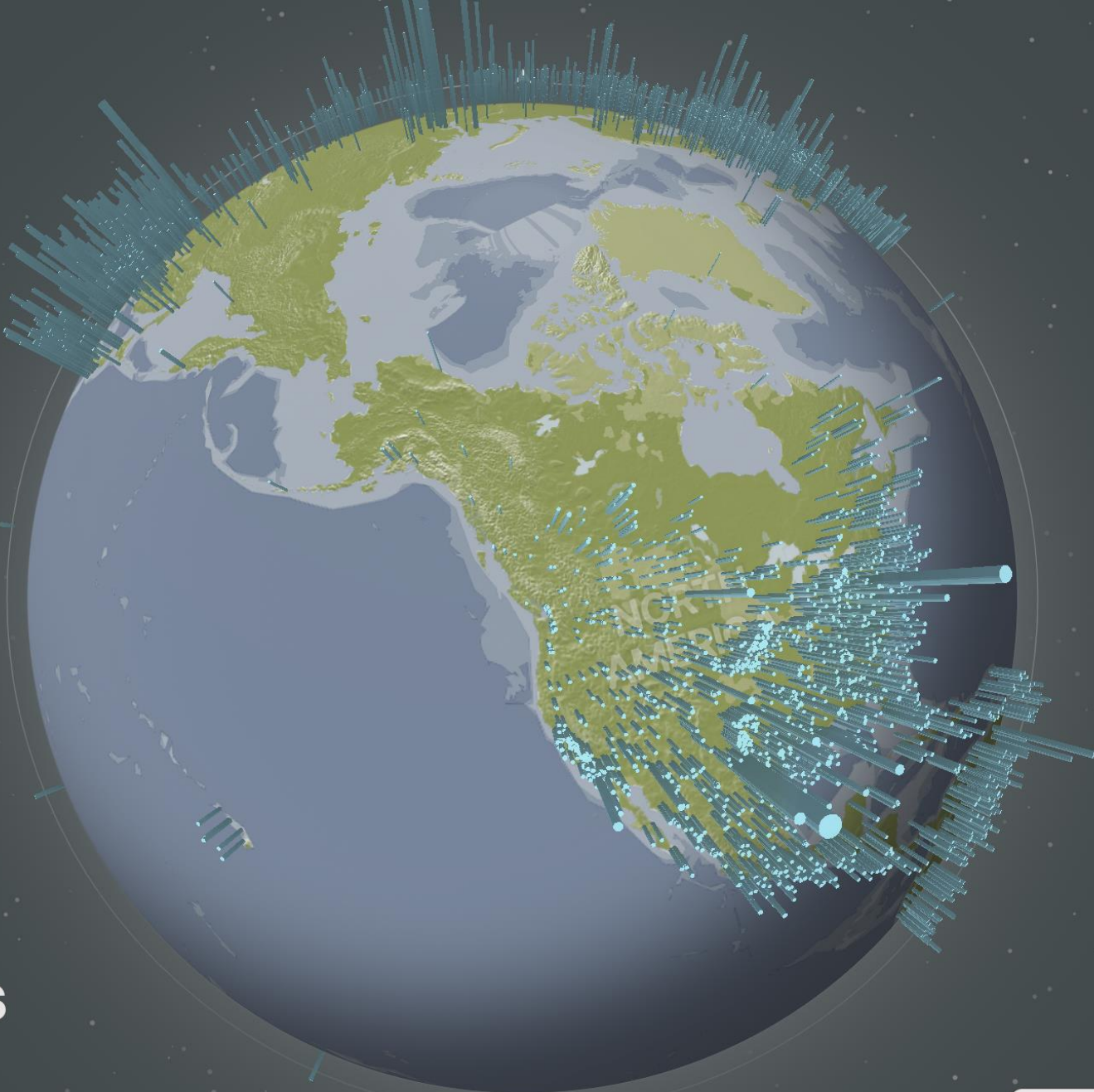




All sectors

Al
Gore

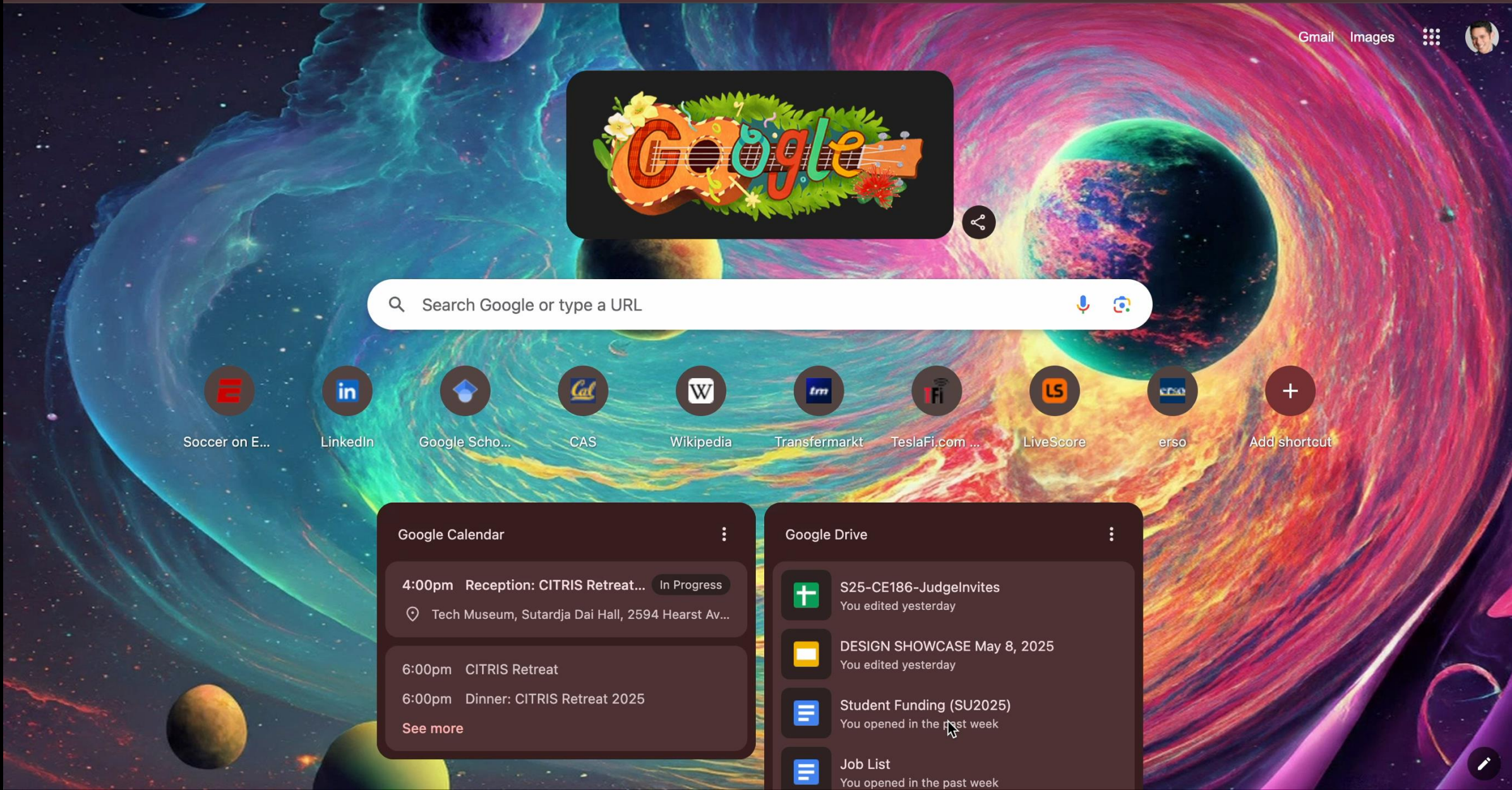
STRATEGY | AL GORE



Independent Greenhouse Gas Emissions Tracking

→ Explore Map

→ Download Data



Gmail Images



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- Google Scho...
- CAS
- Wikipedia
- Transfermarkt
- TeslaFi.com ...
- LiveScore
- erso
- Add shortcut

Google Calendar

4:00pm Reception: CITRIS Retreat... In Progress

Tech Museum, Sutardja Dai Hall, 2594 Hearst Av...

6:00pm CITRIS Retreat

6:00pm Dinner: CITRIS Retreat 2025

See more

Google Drive

S25-CE186-JudgeInvites

You edited yesterday

DESIGN SHOWCASE May 8, 2025

You edited yesterday

Student Funding (SU2025)

You opened in the past week

Job List

You opened in the past week

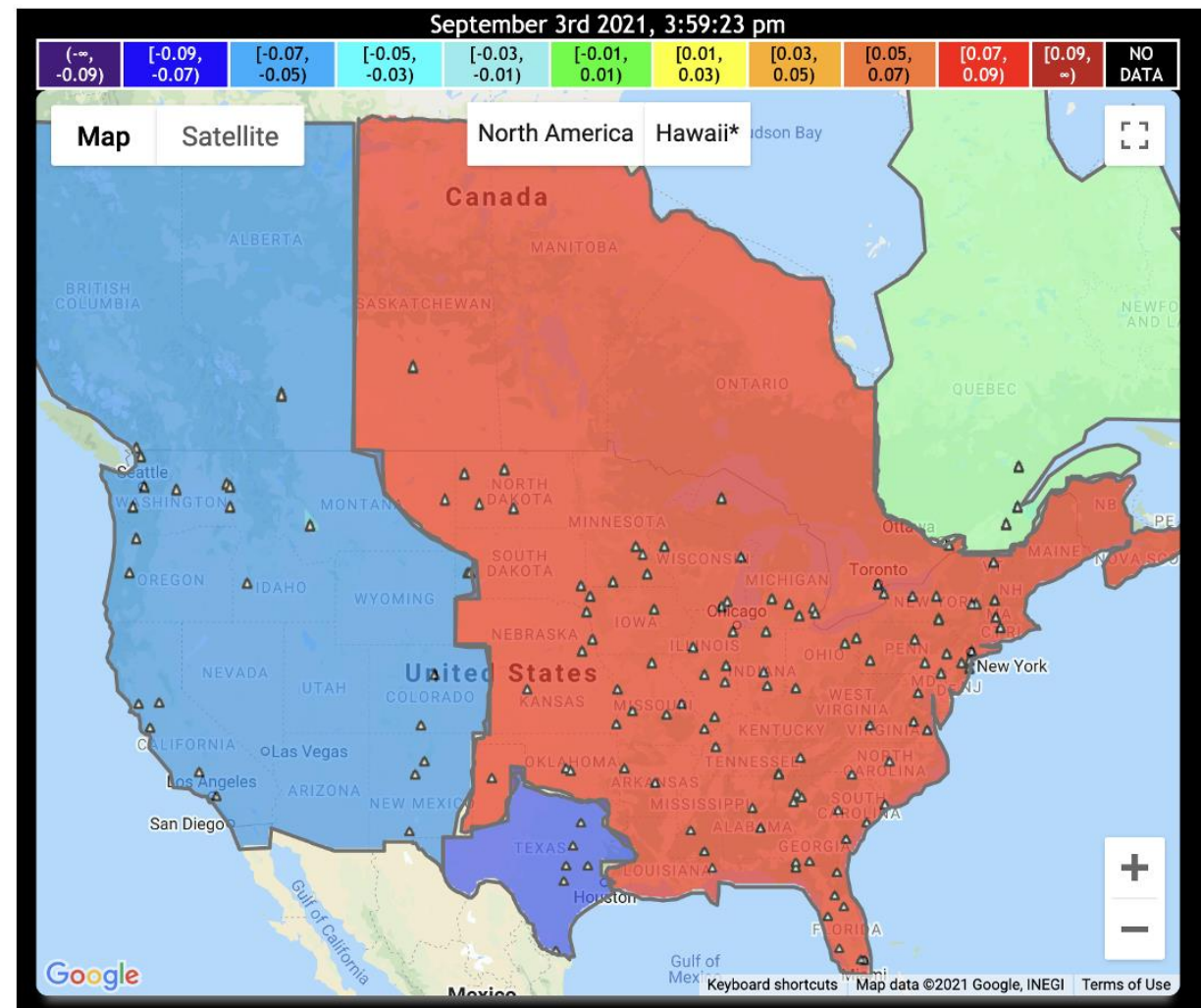
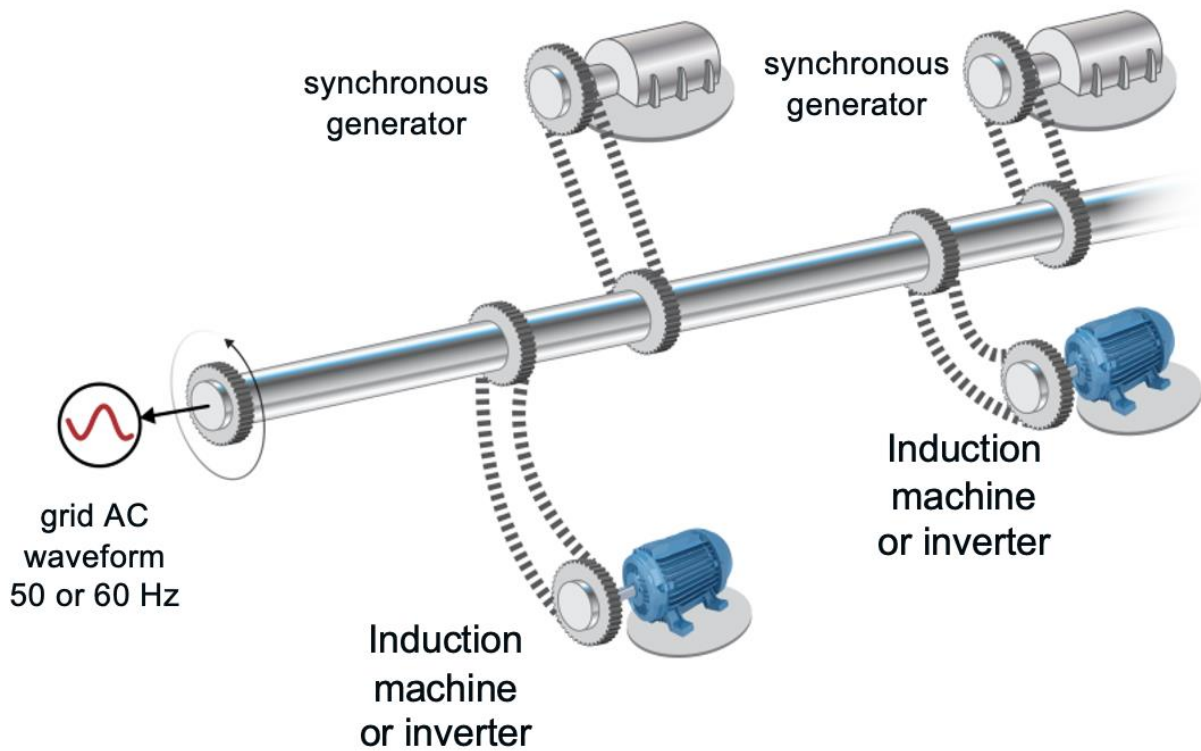


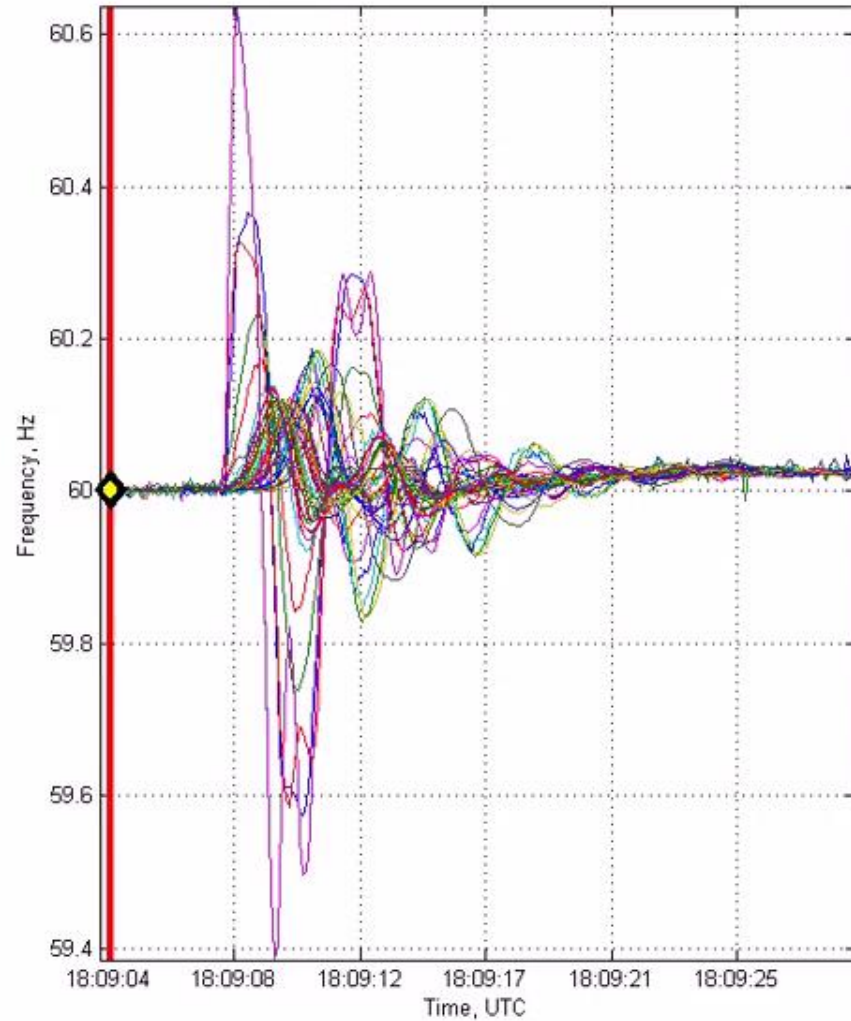




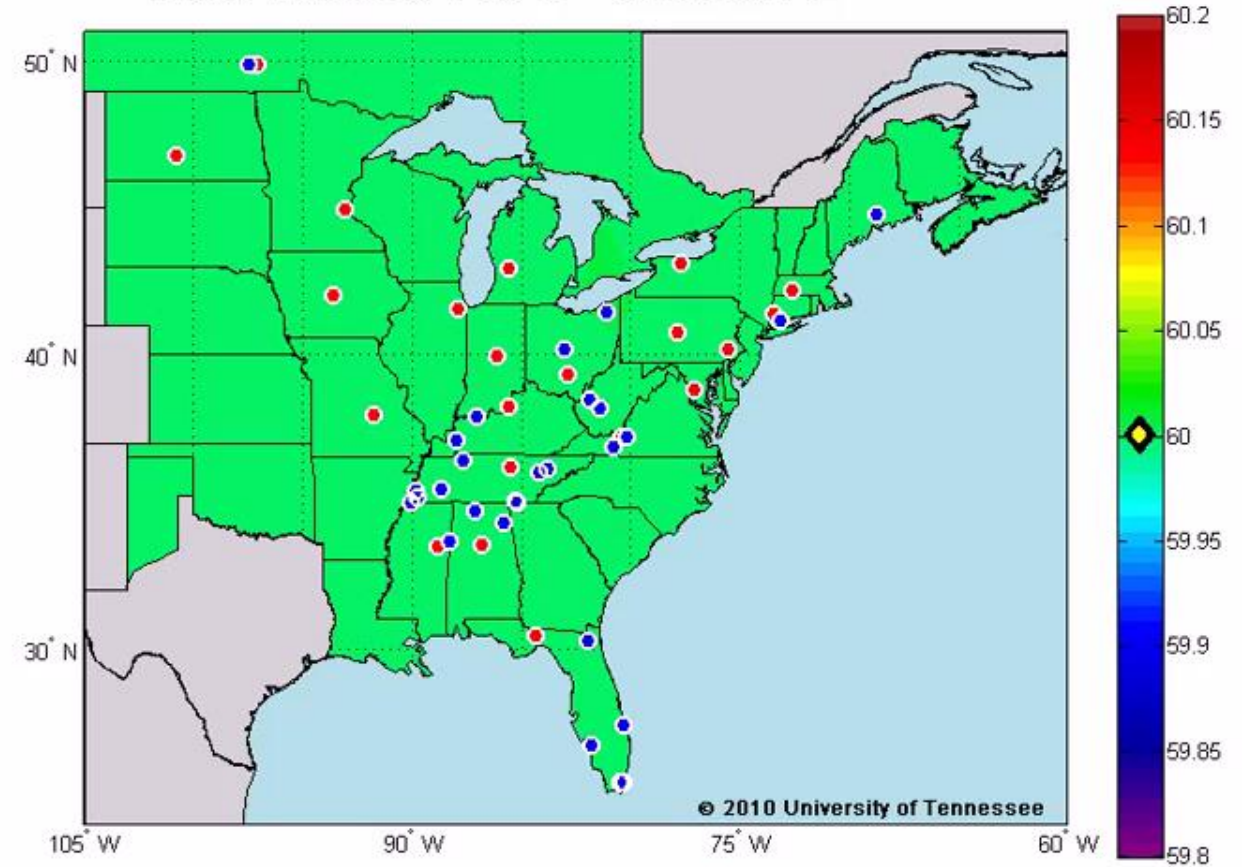








Florida Event Replay with FNET Data [2/26/2008]
Time: 18:09:4.4 UTC 60.0003 Hz



THE UNIVERSITY of
TENNESSEE **UT**

OAK
RIDGE
National Laboratory

CURRENT

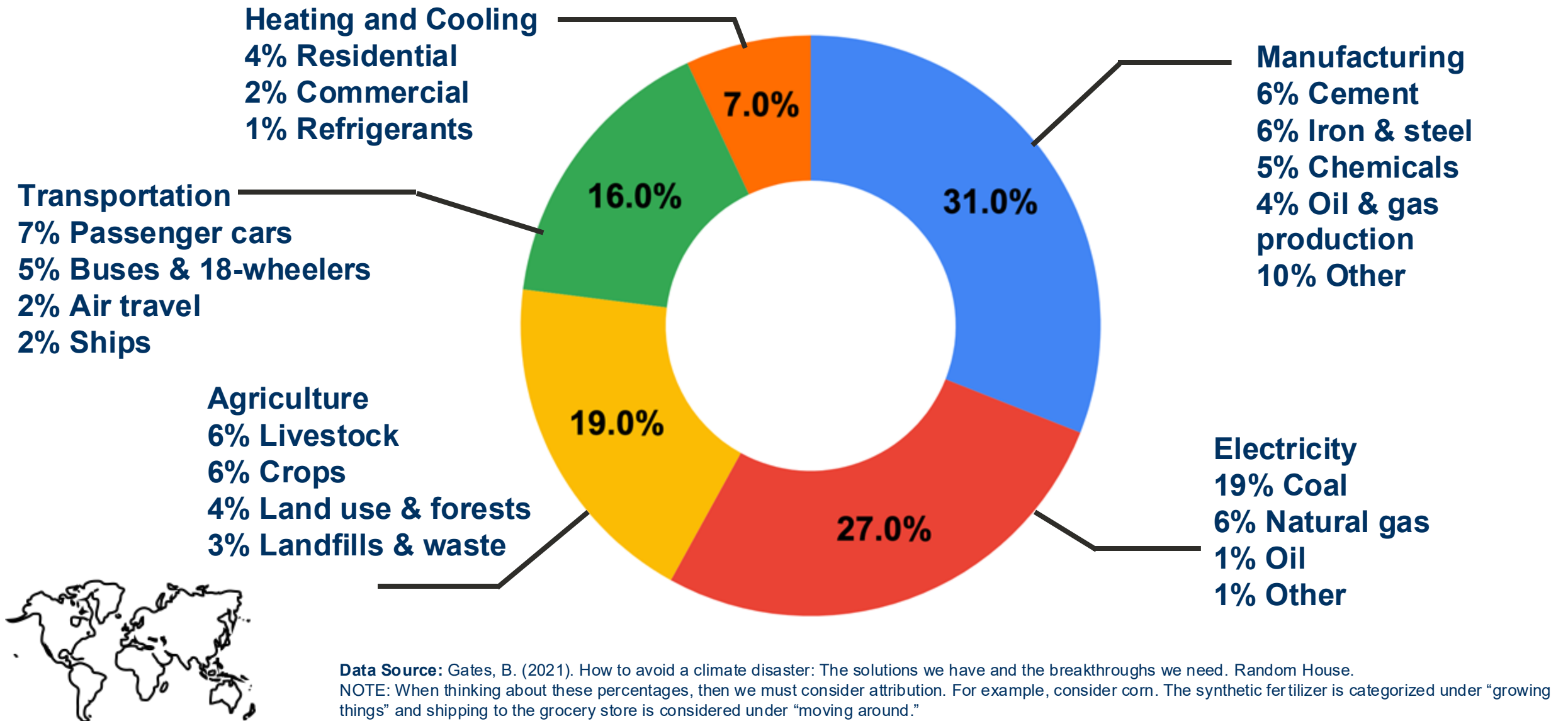
Outline

1. Introduction
2. The Duck Curve Problem
3. Power Grid Impacts of EVs
4. Saving the Grid with SlrpEV
5. Summary

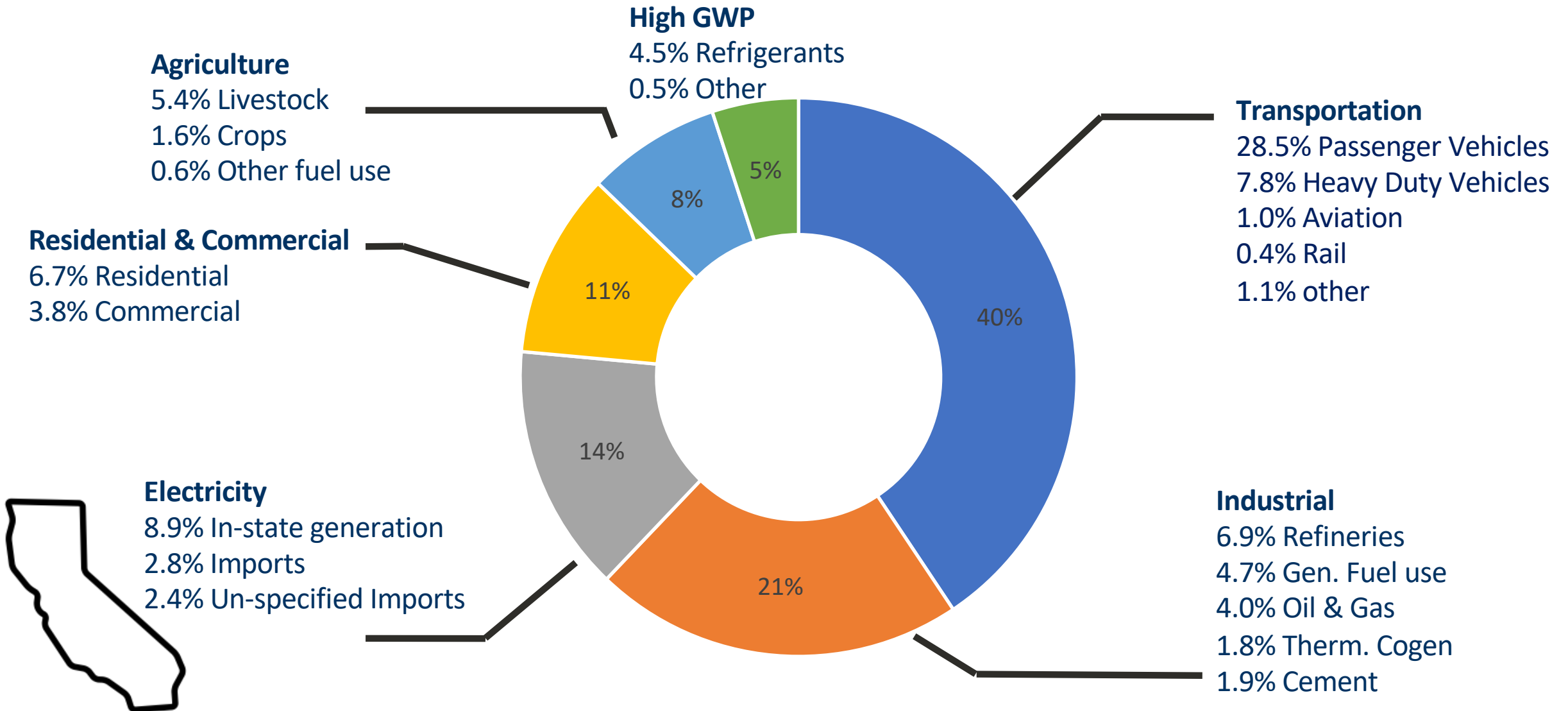
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51 Billion Tons of Greenhouse Gases, Globally



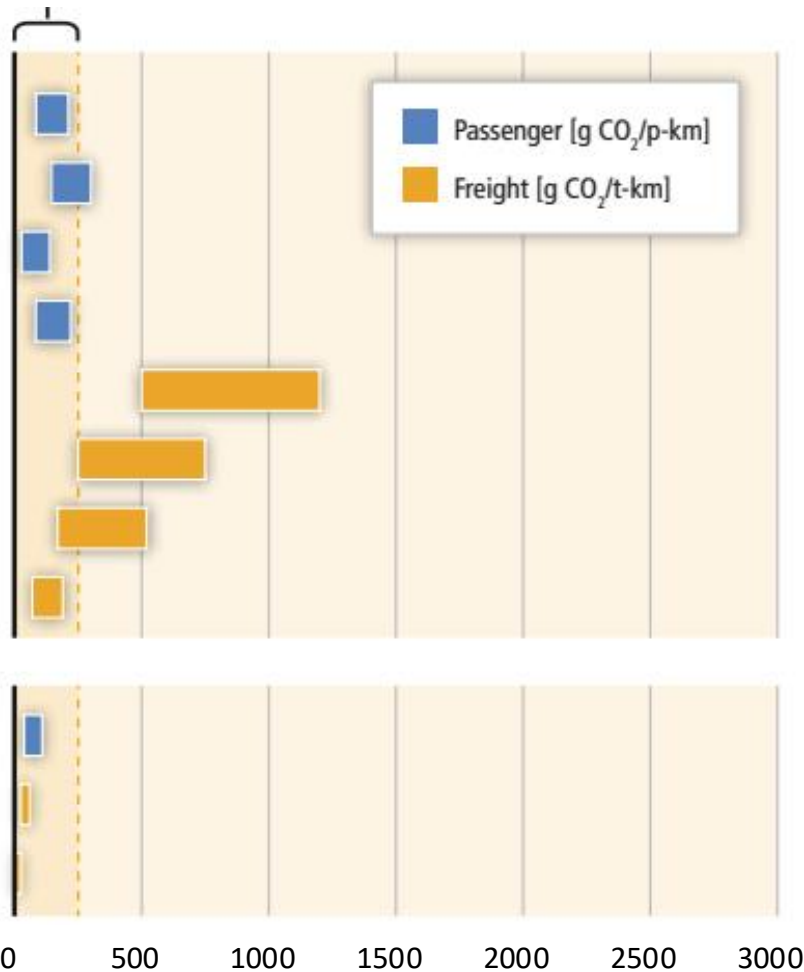
0.42 Billion Tons of Greenhouse Gases, California



Data Source: California Greenhouse Gas Emissions for 2000 to 2019 Trends of Emissions and Other Indicators
https://ww3.arb.ca.gov/cc/inventory/pubs/reports/2000_2019/ghg_inventory_trends_00-19.pdf

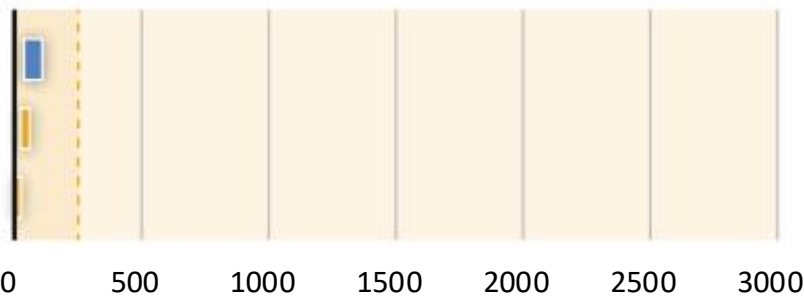
Road

LDV gasoline, diesel, hybrid
LDV Taxi gasoline, diesel, hybrid
Coach, bus, rapid transit
2- and 3-wheel motorbike
LDV commercial (van)
HDV small
HDV medium
HDV large



Rail

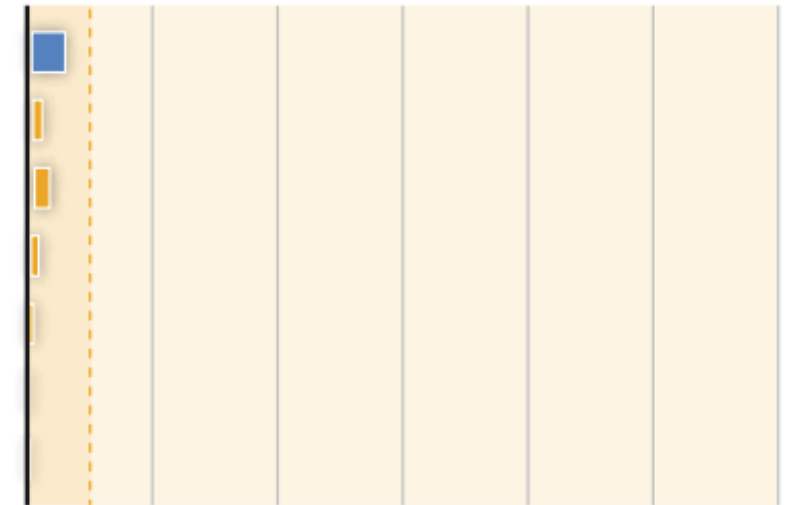
Passenger rail, metro, tram
Diesel freight train
Electric freight train



Direct CO₂ Emissions per Distance
[gCO₂/km]

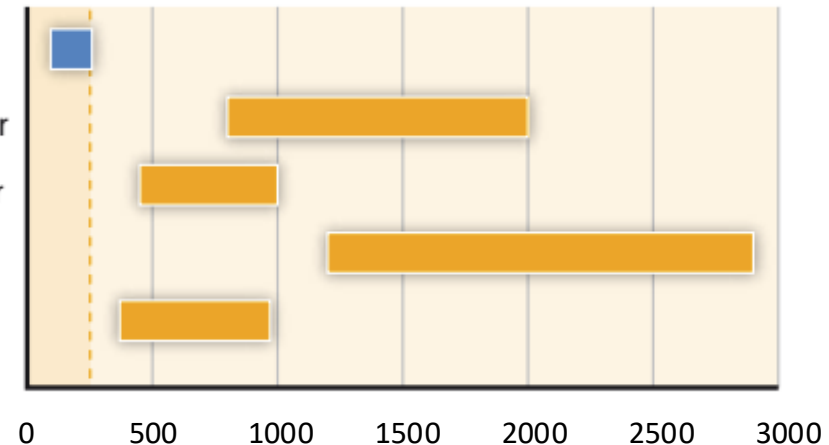
Waterborne

Passenger ferry
Barge
Roll-on, roll-off ferry
Container ship - coastal
Container ship - ocean
Bulk carrier - ocean
Bulk tanker - ocean



Air

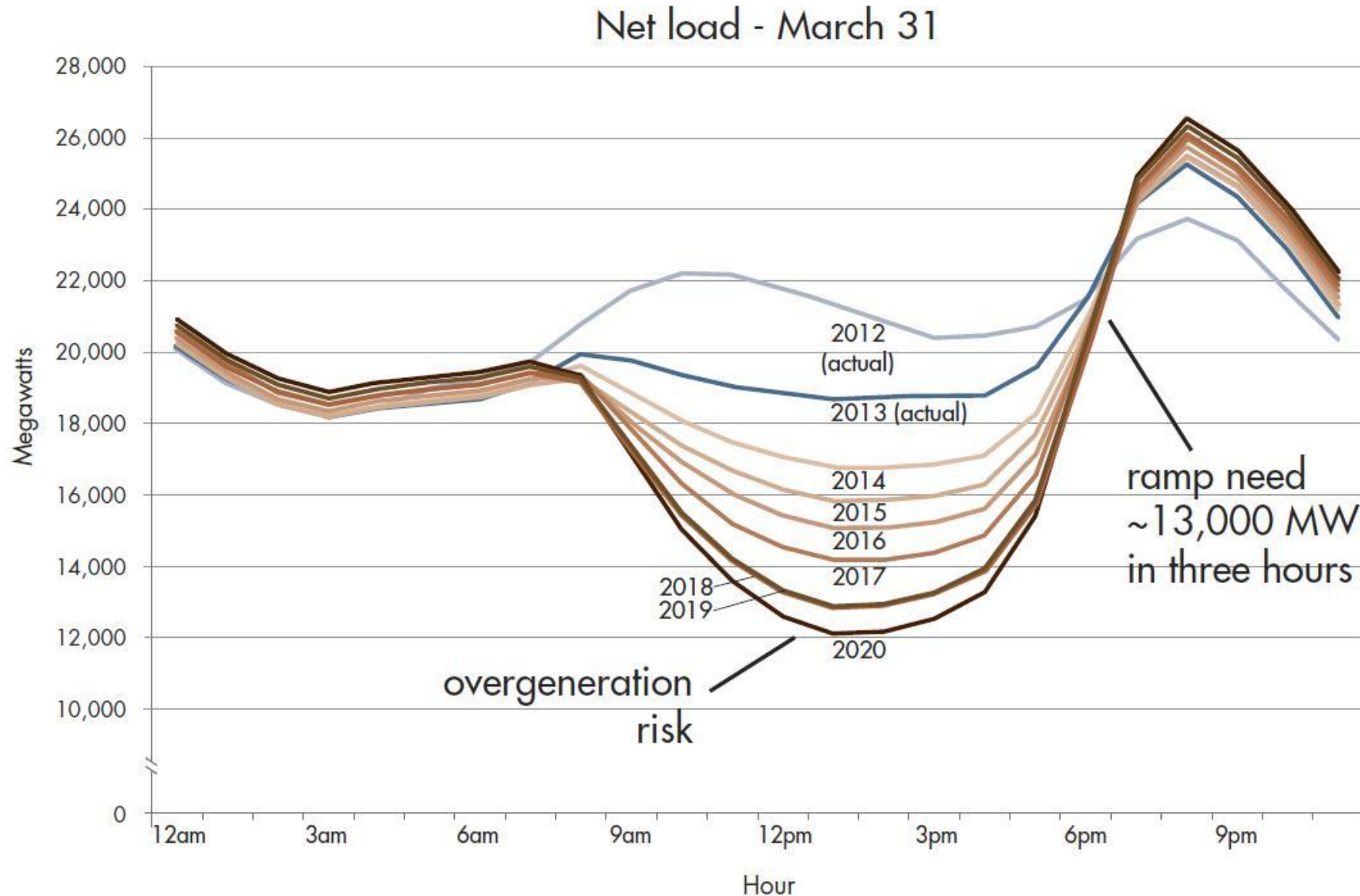
Passenger aircraft
Short-haul bellyhold in passenger
Long-haul bellyhold in passenger
Short-haul cargo aircraft
Long-haul cargo aircraft



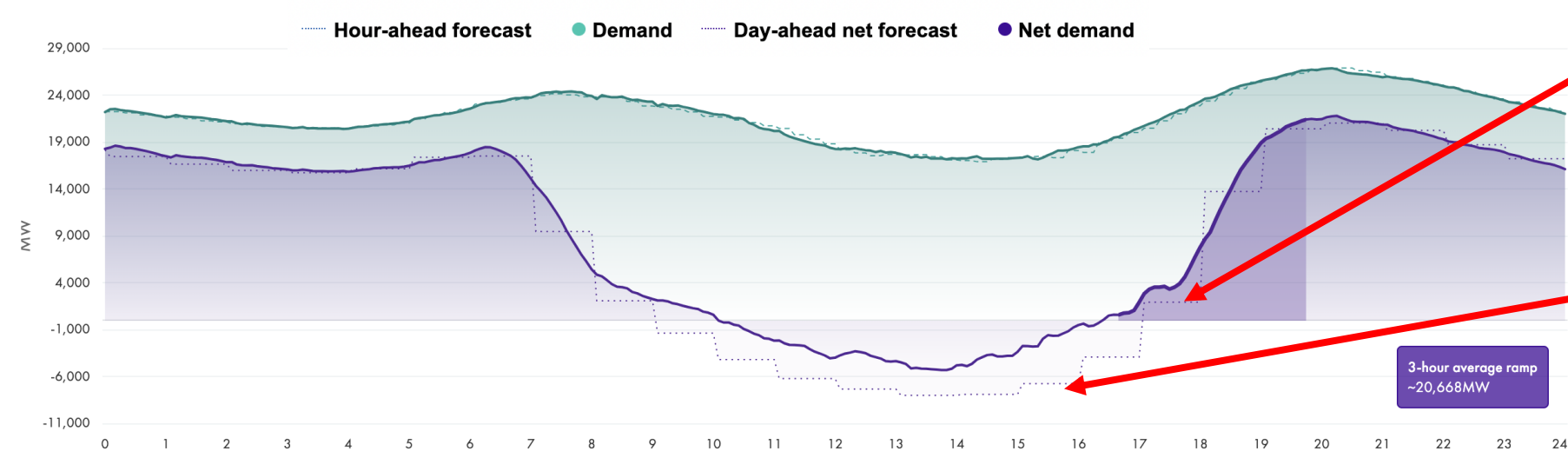
Direct CO₂ Emissions per Distance
[gCO₂/km]

The California Example: Duck Curve

The duck curve shows steep ramping needs and overgeneration risk



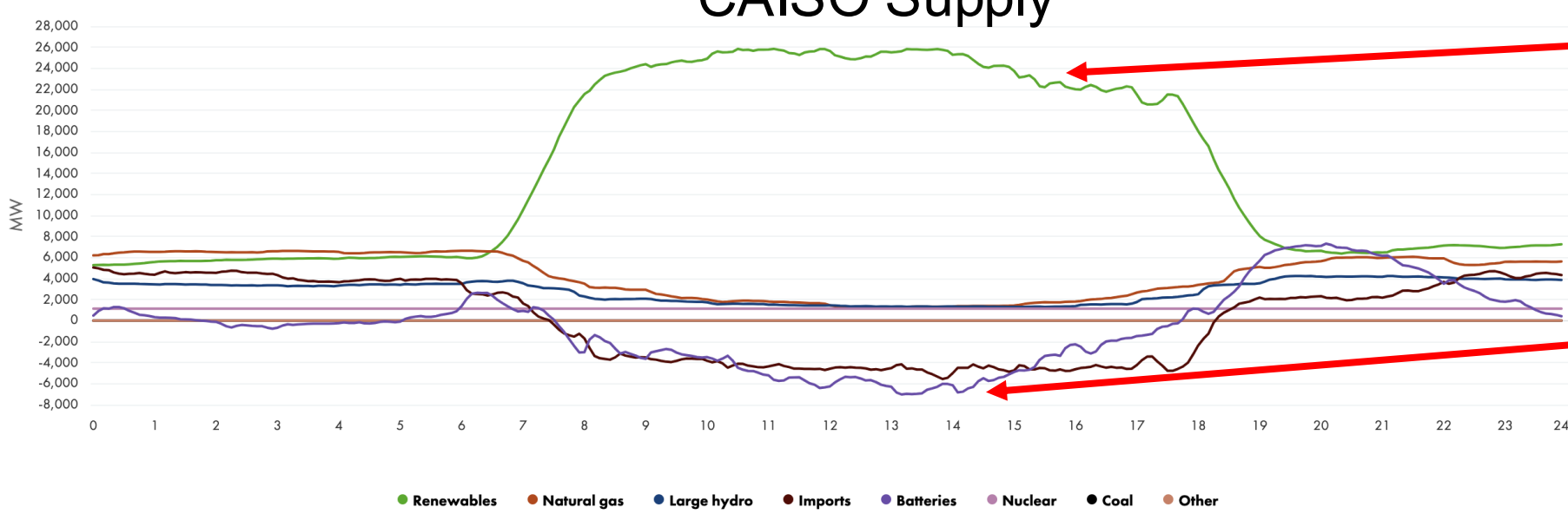
CAISO Electricity Demand



Over 20,000 MW
ramp in 3 hours,
net demand!

Negative net demand!

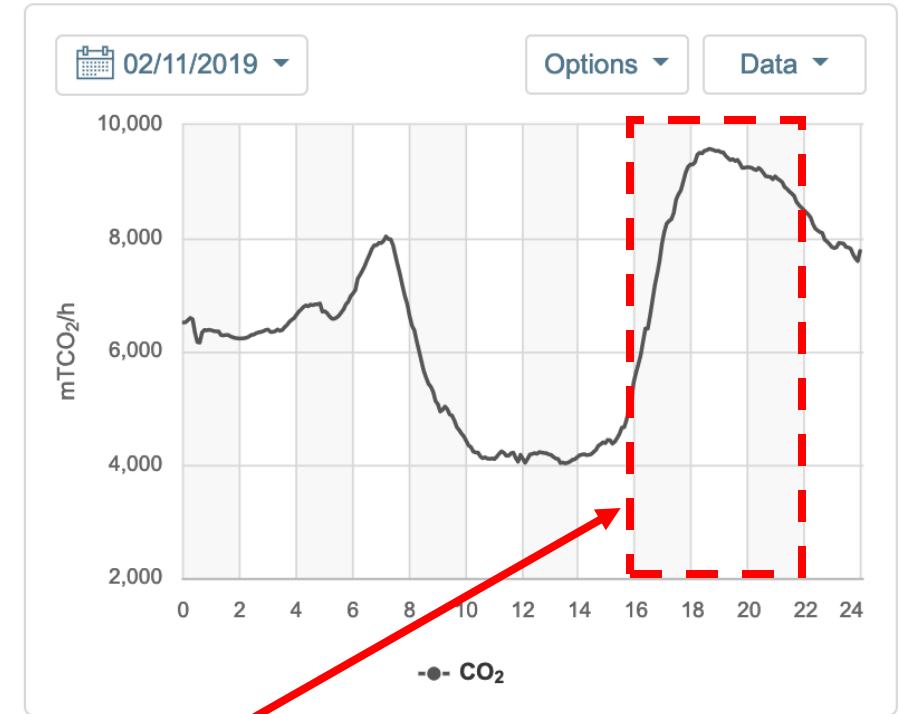
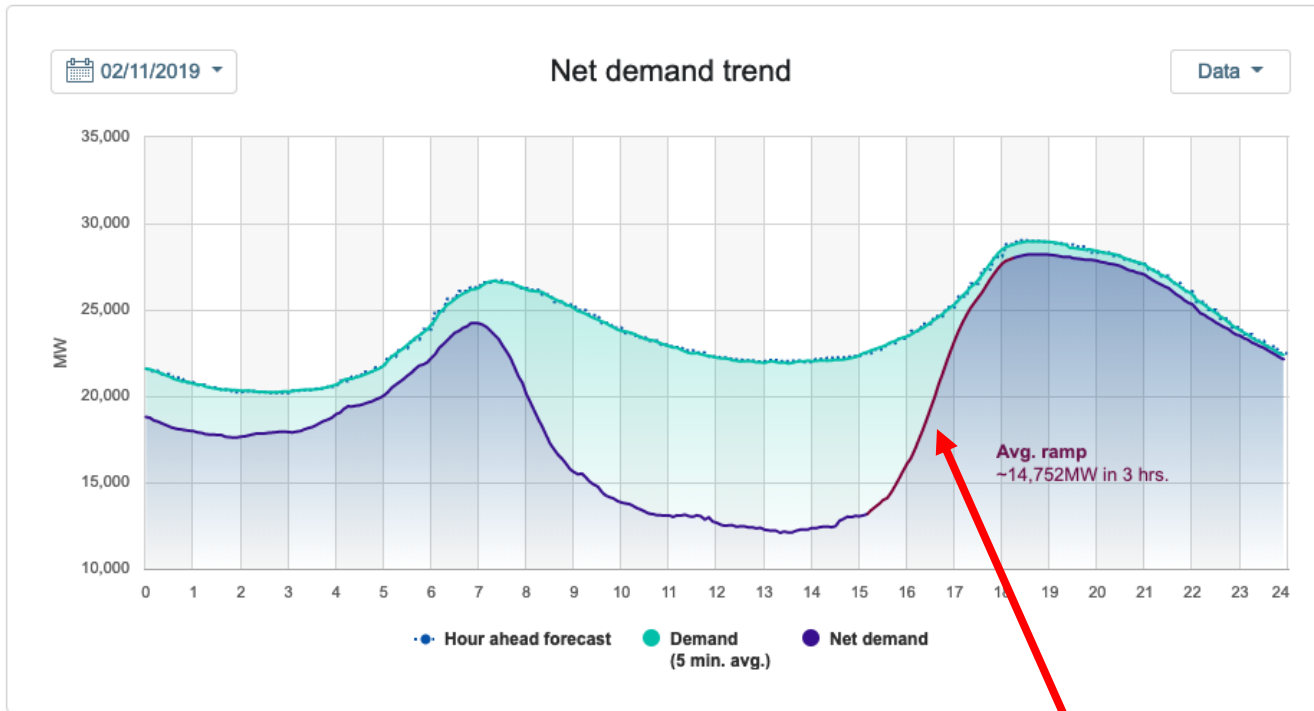
CAISO Supply



Solar curtailment

Max out energy exports,
charge batteries

The Evening Charging Problem



Current challenge, potentially exacerbated by EV penetration:

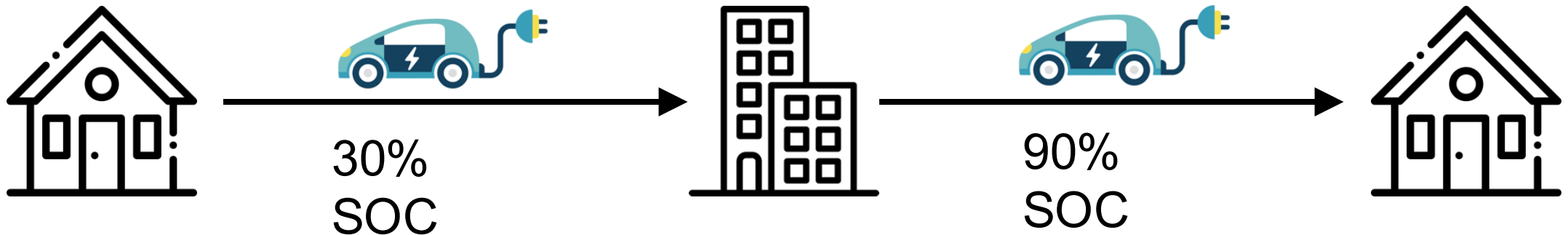
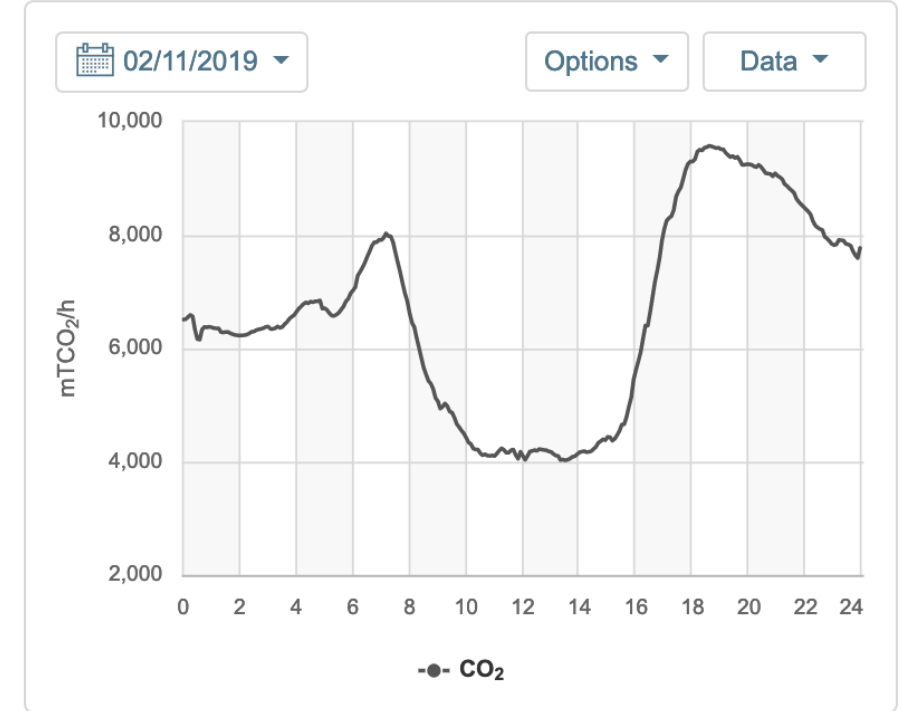
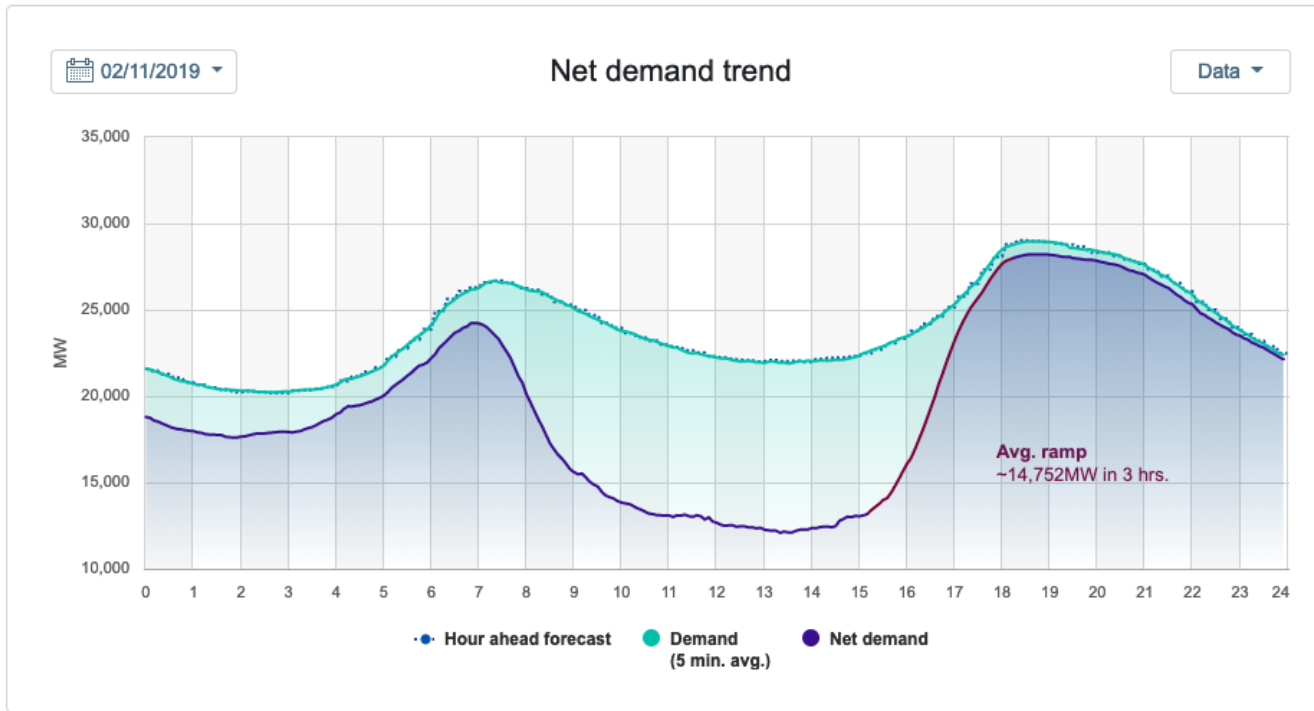
- Ramp rates increase, as folks return home and charge EVs
- CO₂ emissions are typically highest in evening

Opportunity:

- EV charging can be controlled



The Evening Charging Problem



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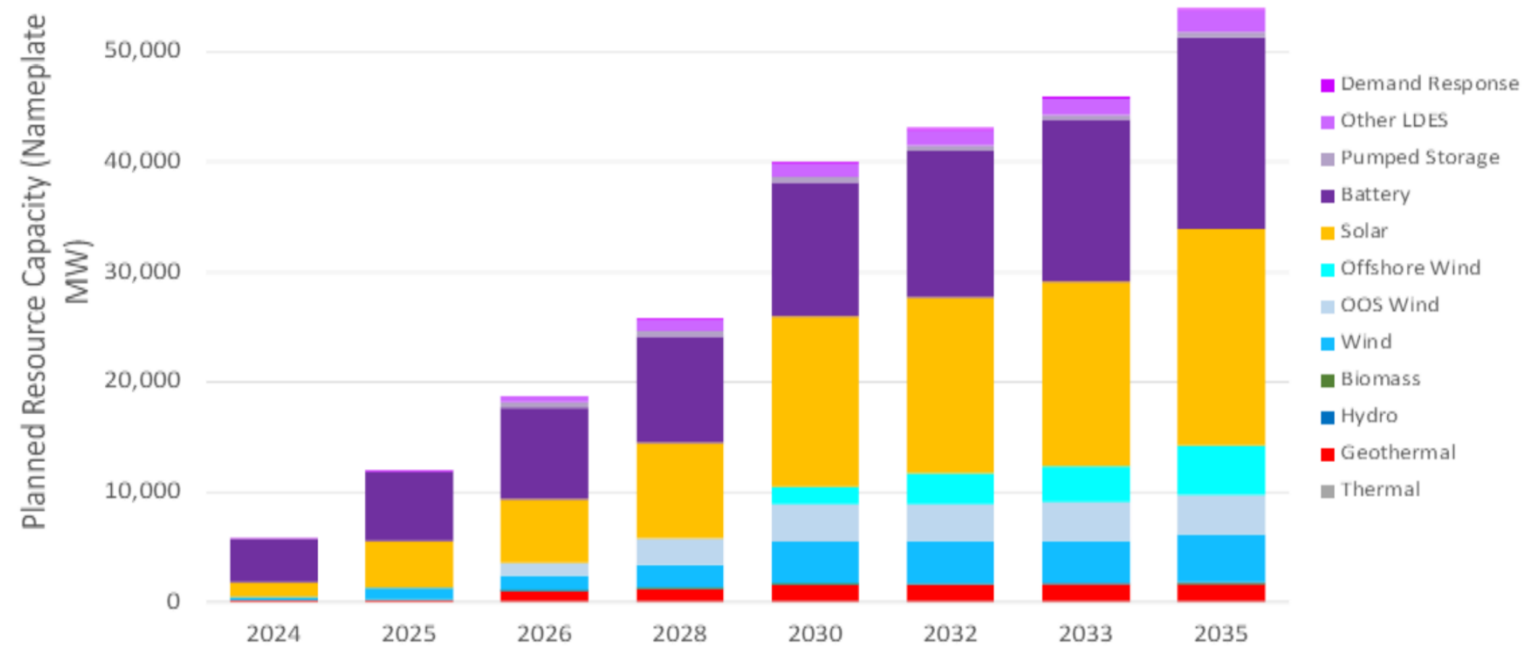
Increased Demand, Planned Supply

- Transportation accounts for more than 10% of projected electricity load growth.



Source: The Brattle Group, [Electricity Demand Growth and Forecasting in a Time of Change](#), May 2024

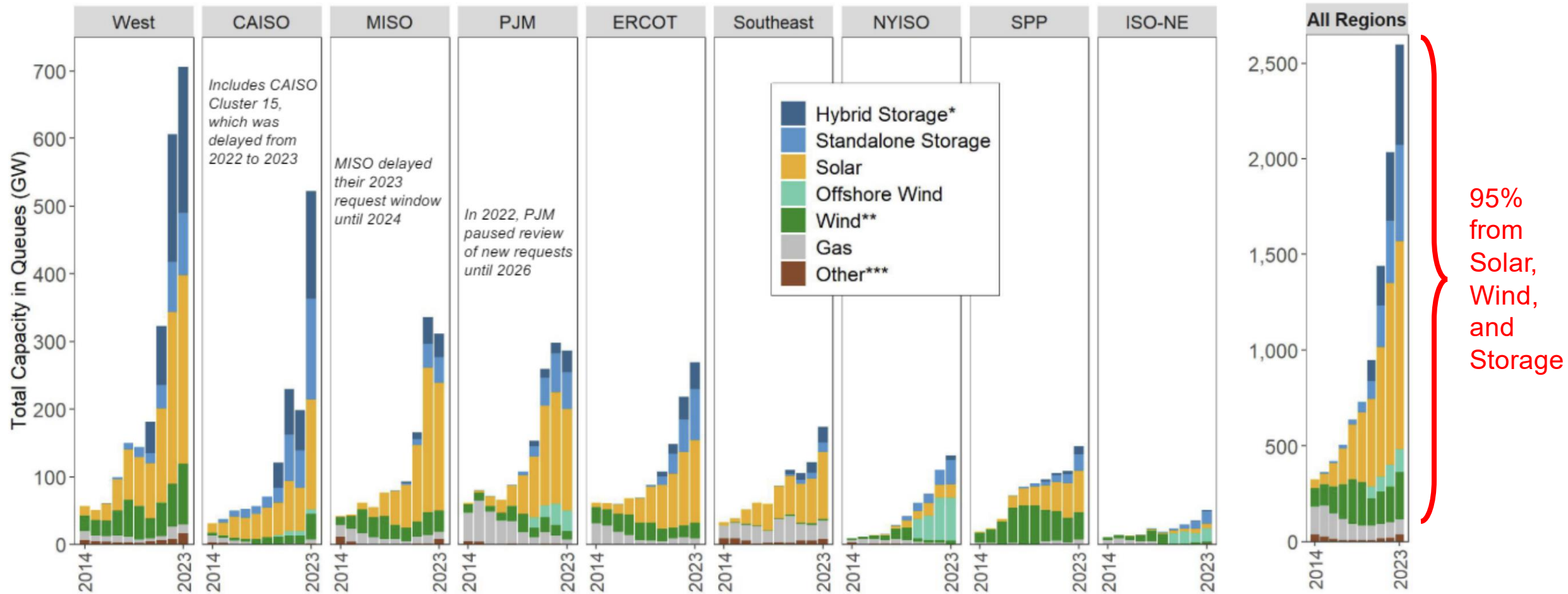
- California plans to add 52 GW of new clean energy resources through 2035, about 4.3 GW per year
- Solar & Batteries comprise over 60% of planned capacity



Source: CPUC, [Integrated Resource Planning \(IRP\) Proposed 2023 Preferred System Plan \(PSP\) and 2024-2025 Transmission Planning Process Portfolios Analysis](#), October 2023.

Solar, Wind, & Storage Dominate the Queue of New Electricity Generation

Total Active Capacity in Interconnection Queues across the U.S.

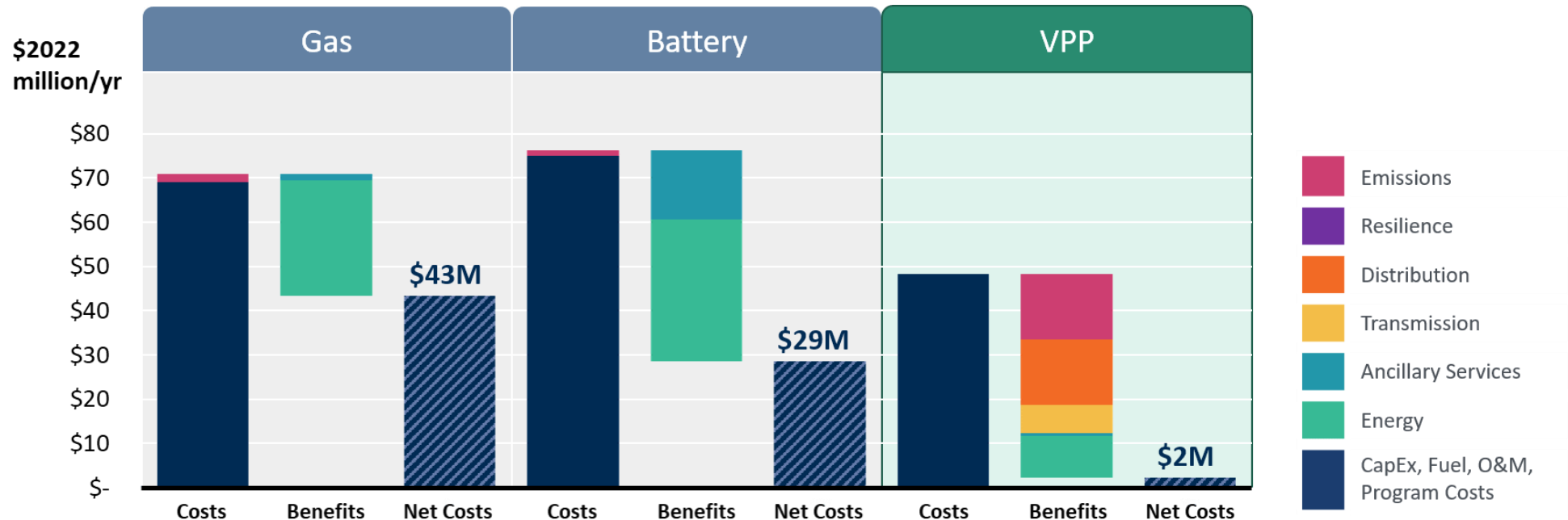


Only ~20% of projects (and 14% of capacity) requesting interconnection from 2000-2018 reached commercial operations by the end of 2023 🤖🤖🤖

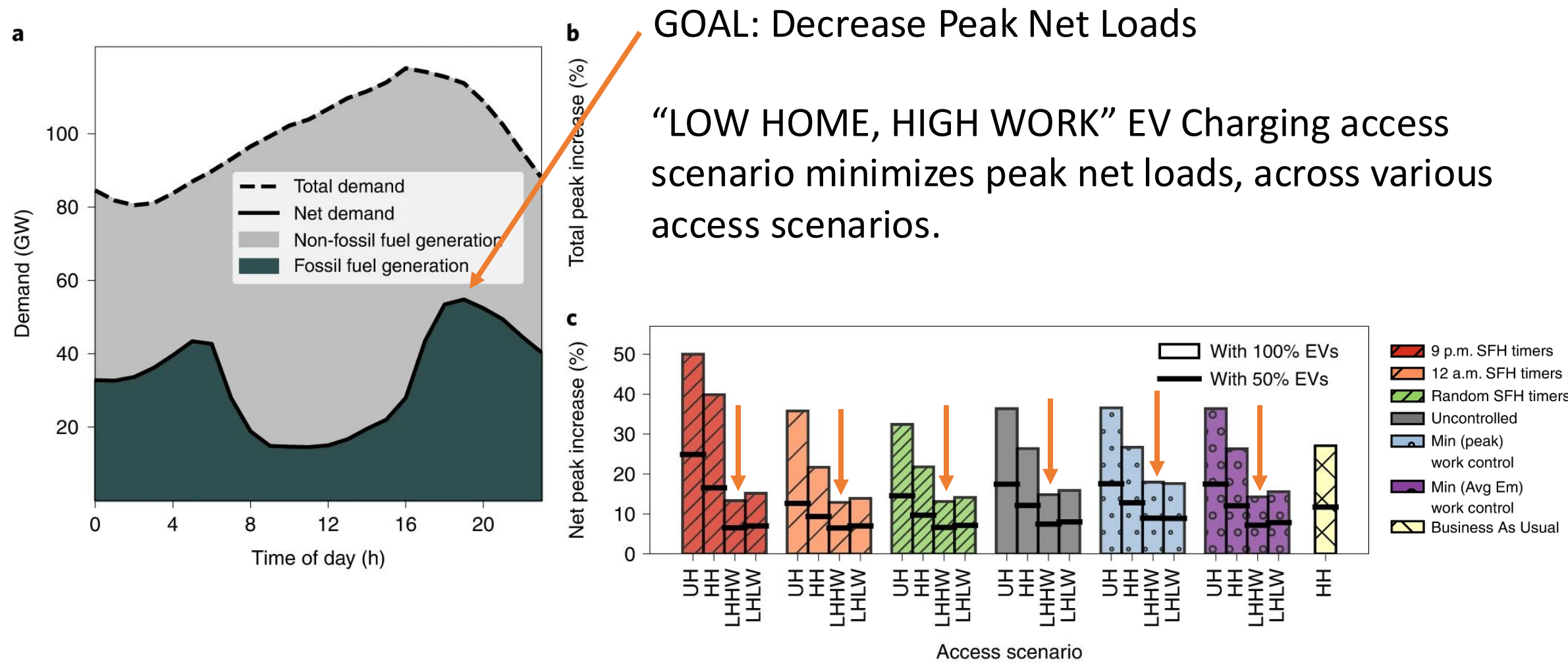
Source: LBNL, [Queued Up: 2024 Edition: Characteristics of Power Plants Seeking Transmission Interconnection As of the End of 2023](#), April 2024

Demand Flexibility via Virtual Power Plants (VPPs)

- Many categories of electric demand are flexible: EVs, HVAC, water heaters, etc.
- Many categories of electric demand are controllable.
- VPP: A portfolio of distributed energy resources that are actively controlled to provide benefits to various stakeholders, e.g. all EVs in California.
- Avoid interconnection delay, lower cost!

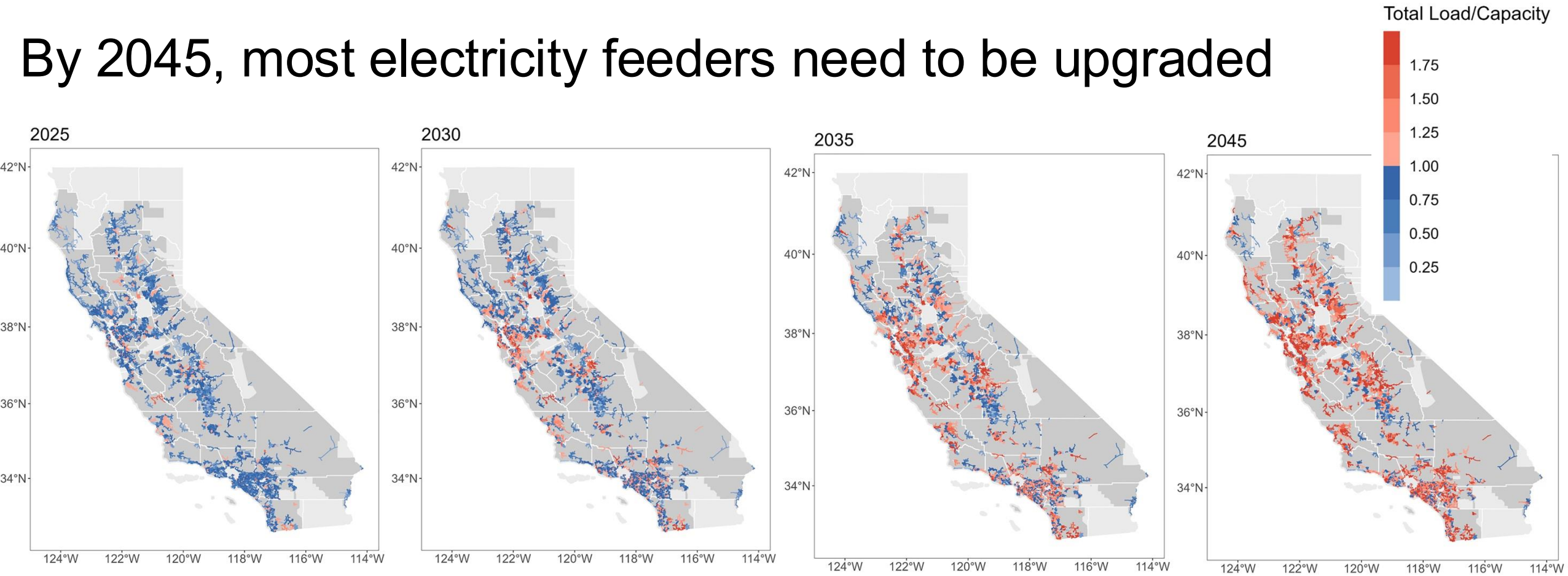


High Work Charging Decreases Peak Net Loads



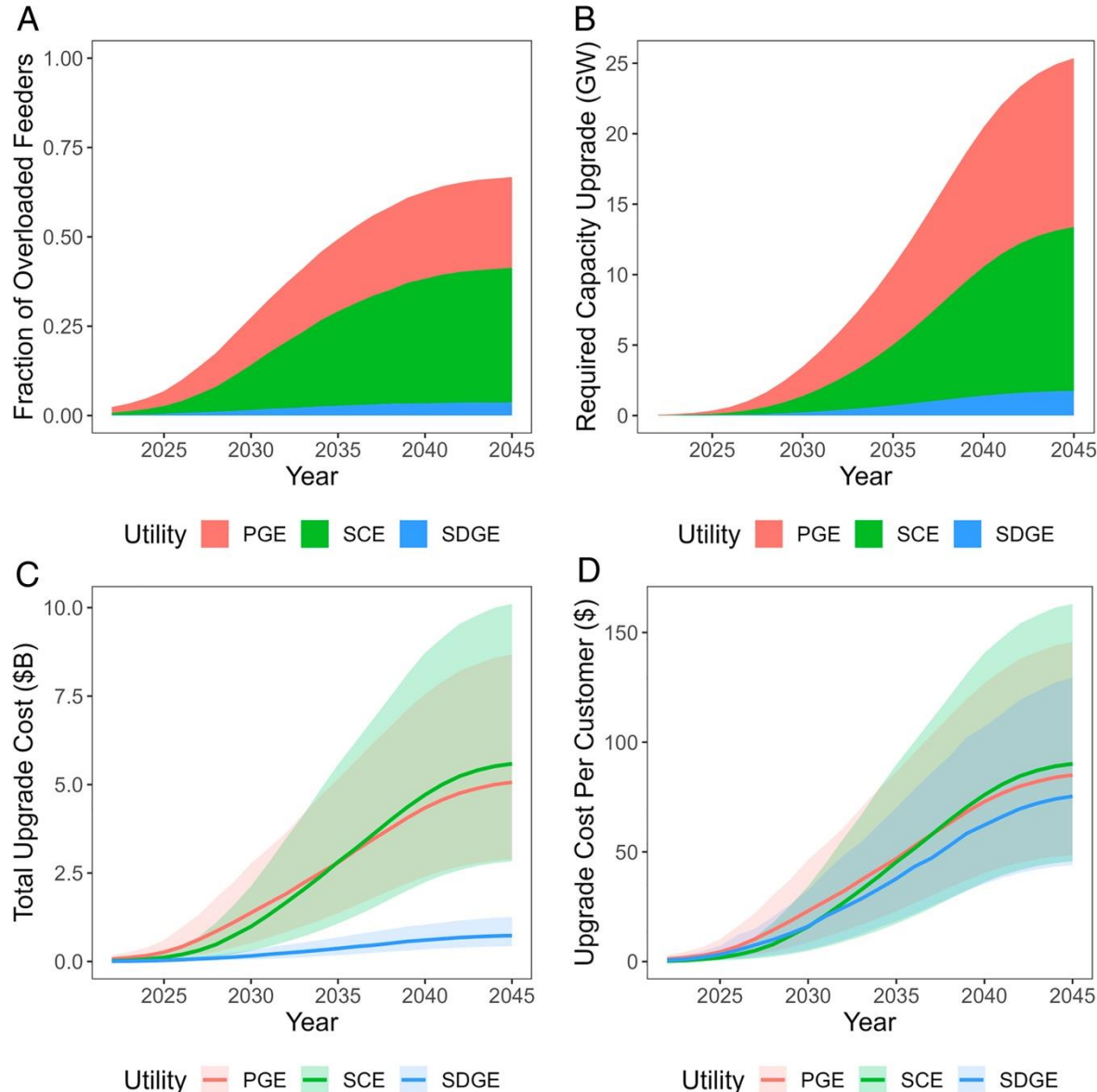
Source: S. Powell, G. V. Cezar, L. Min, et al. *Charging infrastructure access and operation to reduce the grid impacts of deep electric vehicle adoption*. Nature Energy (2022). <https://doi.org/10.1038/s41560-022-01105-7>

By 2045, most electricity feeders need to be upgraded



Source: Y. Li, A. Jenn, "Impact of electric vehicle charging demand on power distribution grid congestion," *Proceedings of the National Academy of Sciences*, April 2024. DOI: [10.1073/pnas.2317599121](https://doi.org/10.1073/pnas.2317599121). Edited by **SJM**.

67% of feeders are overloaded by 2045



- Southern California Edison have greater quantity of overloaded feeders, but...
- Pacific Gas & Electric has greater magnitude of overloading
- Although upgraded infrastructure puts upward pressure on price, increased load demand places downward pressure on price. Net effect? Close to zero.

Source: Y. Li, A. Jenn, "Impact of electric vehicle charging demand on power distribution grid congestion," *Proceedings of the National Academy of Sciences*, April 2024. DOI: [10.1073/pnas.2317599121](https://doi.org/10.1073/pnas.2317599121). Edited by SJM.

Soo... Scott.... will EVs break the grid?



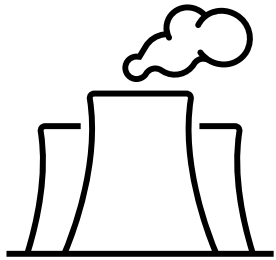
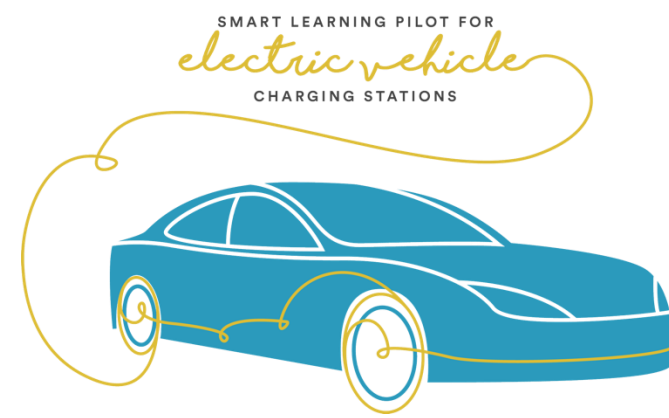
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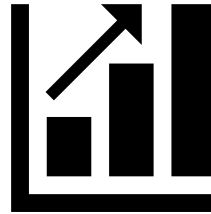
Goal & Objectives

SlrpEV Goal:

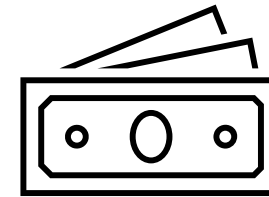
Create next generation of workplace/public EV charging that



decreases emissions



increases facility
operator revenues



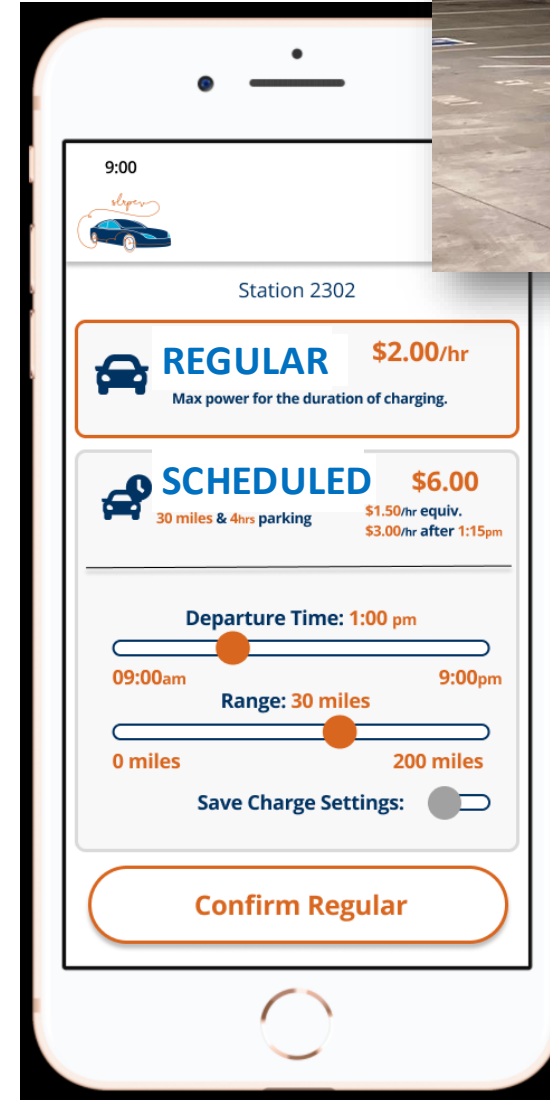
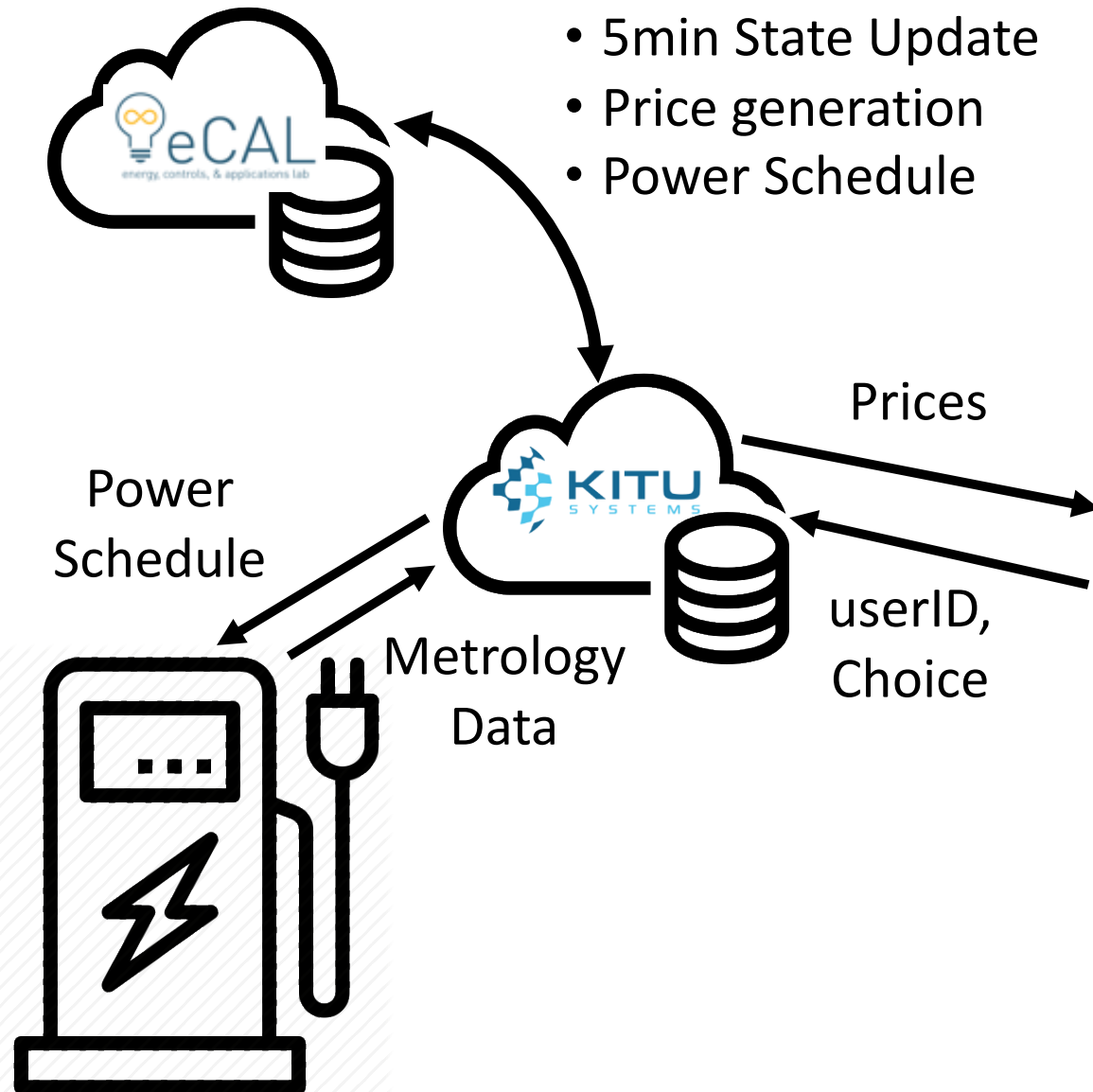
decreases EV
owner costs

Research Objective:

- Optimize price and charging schedule by ***learning user preferences***

Cyber-Physical & Human System

The Cyber & Physical



UC Berkeley
8 Level 2 chargers
(~6.6 kW)

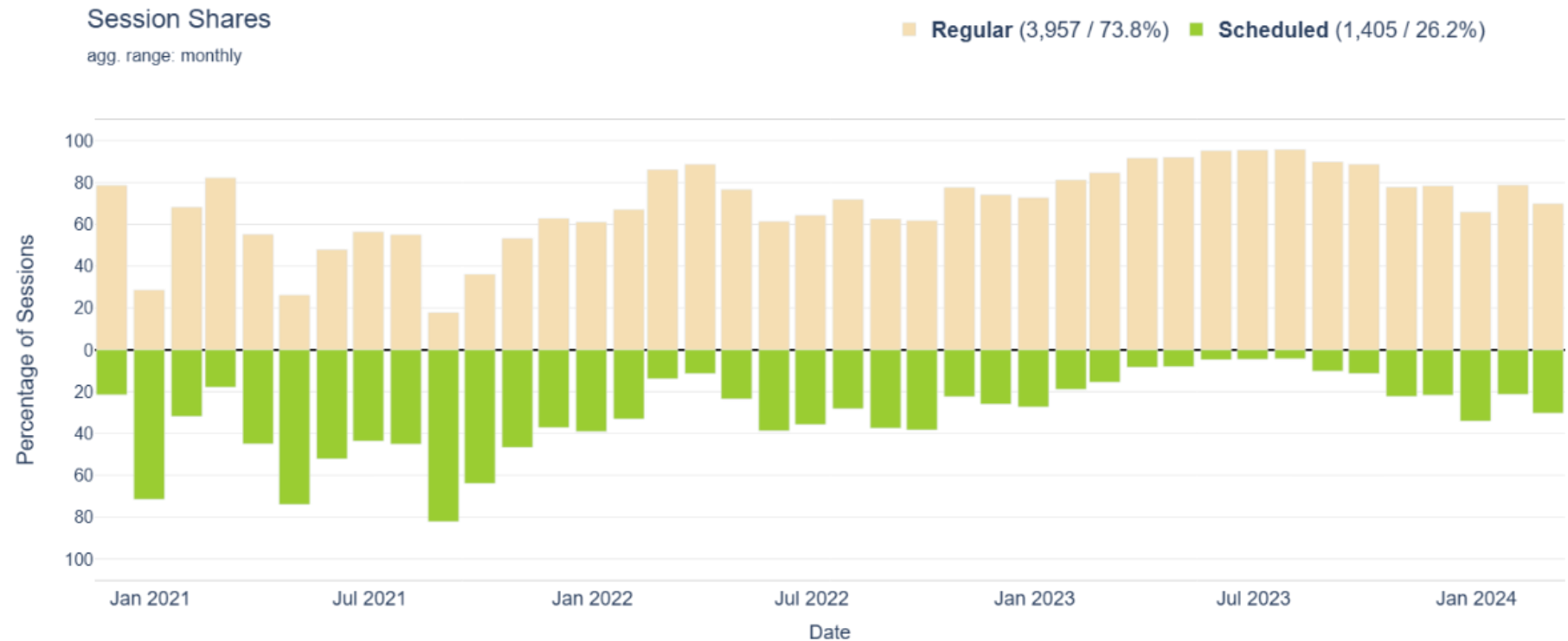
SlrpEV pilots flexible EV charging at the workplace



Smart Learning Pilot for Electric Vehicle Charging Stations

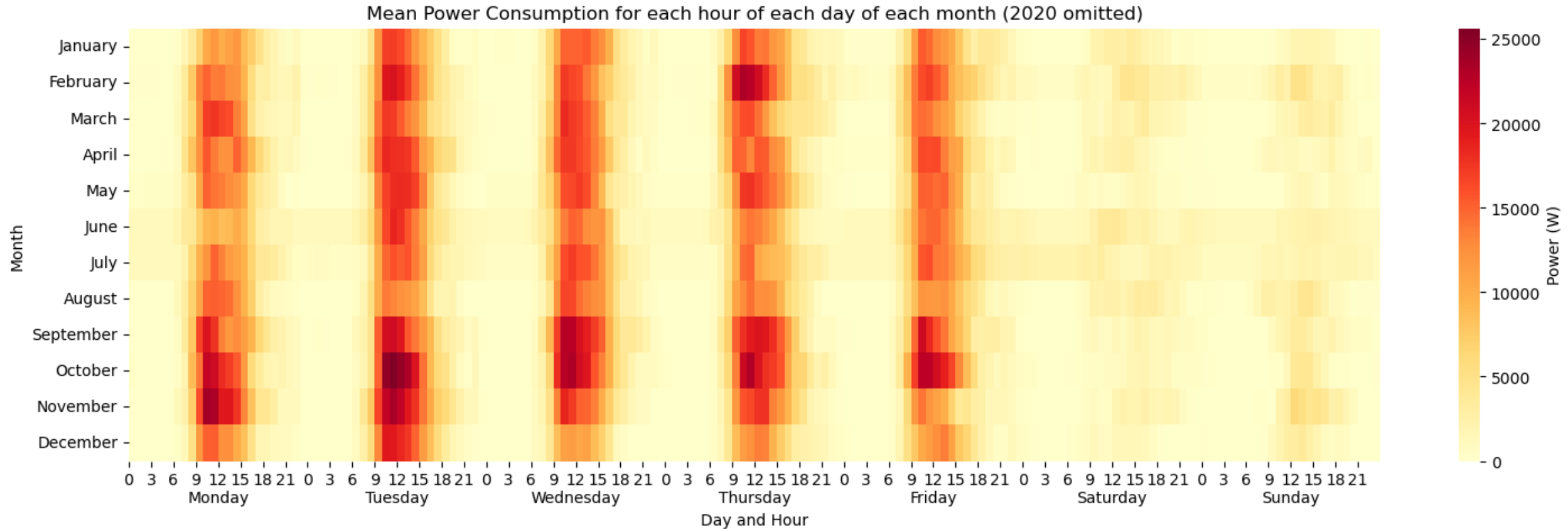
8 level-2 chargers on Berkeley campus (stats updated by 29 Dec 2023)

267 unique users, 4962 sessions, 96046 kWh delivered



Explore our [data dashboard: http://slrpev.fun.yi-ju.me/](http://slrpev.fun.yi-ju.me/)

Seasonality of Electric Power Demand



- Demand concentrated between 10a – 12n on weekdays
- Demand increases during Sept-Nov, Feb-Mar

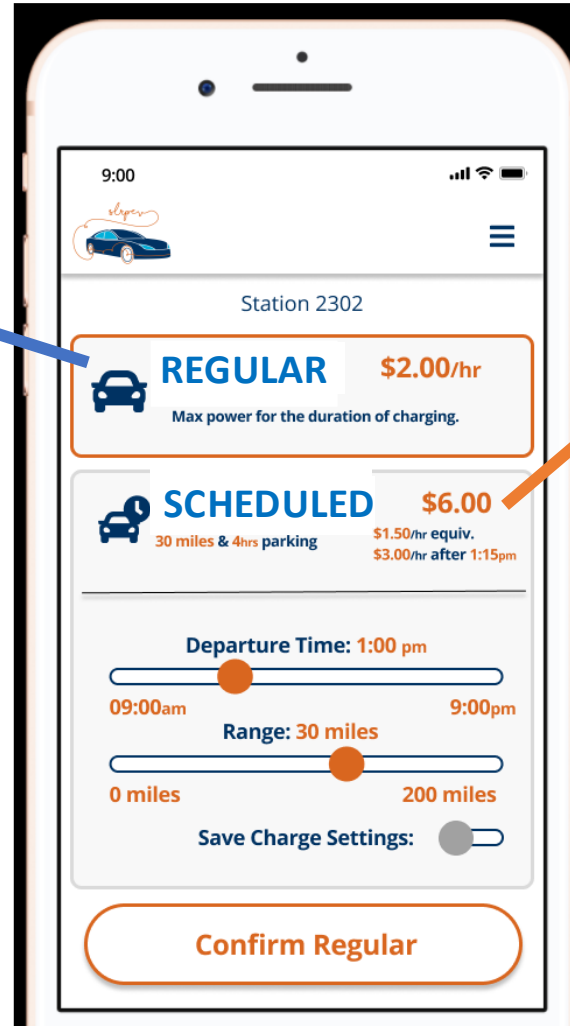
Cyber-Physical & Human System

The Human



REGULAR

- Fixed rate in USD/hr
- Max power until unplug or top-off
- NOTE: charging power is uncontrollable load



SCHEDULED

- User provides departure time & added range
- Total cost fixed *a priori*
- NOTE: charging power is now controllable

If I want to shift load, then how much should you discount SCHEDULED to acquire flexibility?



SAVE MONEY AT SLRPEV!

Discrete Choice Model – How to model human behavior

Daniel McFadden
2000 Nobel Prize in
Economic Sciences



$$U_j = \beta_j^T \mathbf{z}_j + \gamma_j^T \mathbf{w}_j + \beta_{0j} + \varepsilon_j$$

where

U_j : Utility of j-th alternative, $j \in \{\text{asap}, \text{flex}\}$

β_j : Parameters of controlled attributes

\mathbf{z}_j : Controlled attributes (e.g. prices)

γ_j : Parameters of UN-controlled attributes

\mathbf{w}_j : Uncontrolled attributes (e.g. arrival time)

β_{0j} : Alternative specific constant

ε_j : Undefined errors

Logit Model

Assuming “perception” errors ε_j have i.i.d. Extreme Value Distribution, the prob. of choosing j-th alternative is

$$\begin{aligned} \Pr(\text{alt } j \text{ chosen}) &= \Pr\left(\cap_{j \neq i} (U_j > U_i)\right) \\ &= \frac{e^{V_j}}{\sum_{i=1}^J e^{V_i}} = sm(V) \end{aligned}$$

where $V_j = \beta_j^T \mathbf{z}_j + \gamma_j^T \mathbf{w}_j + \beta_{0j}$



Optimizing Price & Power

Expected Cost Minimization Problem

$$\min_{z,u} \sum_j \Pr(J = j|z) h_j(z, u)$$

subject to: linear functions of (z,u)

where z is incentive control, u is direct control, and $h_j(z, u)$ is bi-convex in (z, u) .

Compact Form

$$\min_{z,u} v^T h(z, u)$$

where $v_j = sm(\Theta_j z)$, $h = [h_{flex}(z, u) \quad h_{asap}(z, u)]^T$

Q: How to effectively and efficiently find solutions?

A: Re-formulate into multi-convex problem

Optimize Price

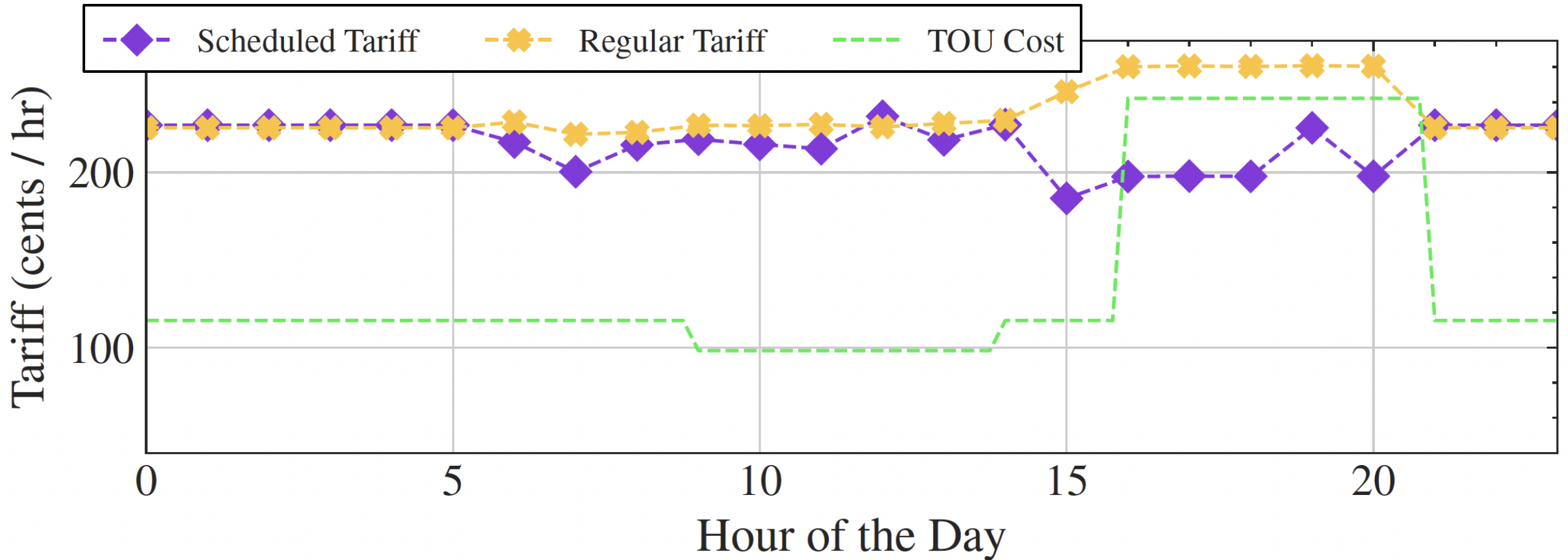
Maximize:

Net revenue

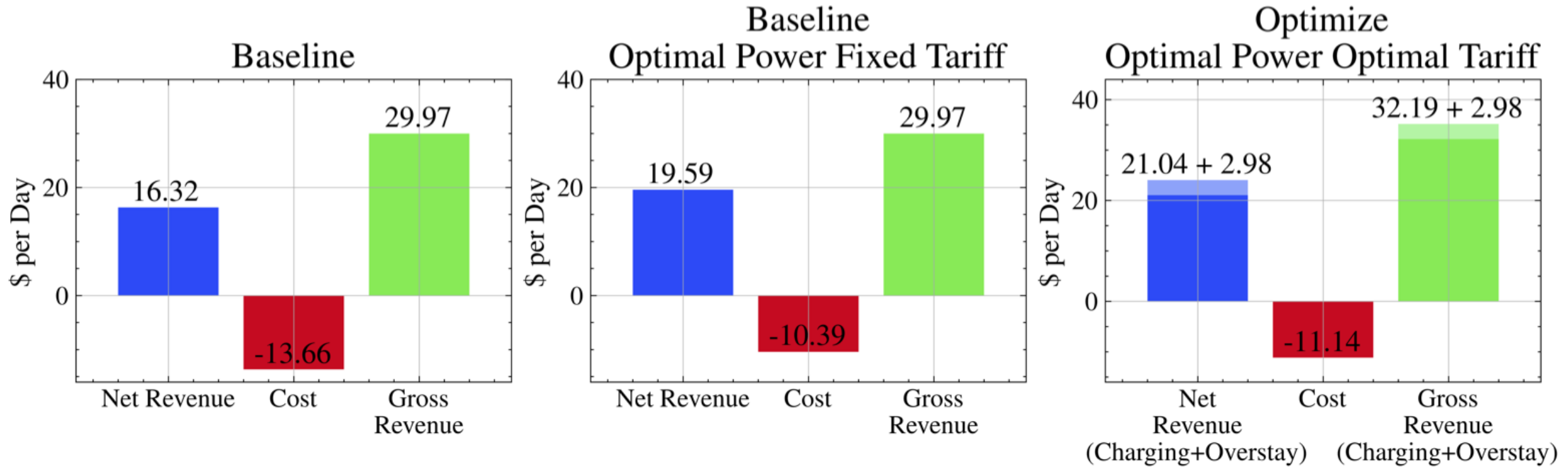
$$= \text{Service Revenue} - \text{Utility Cost}$$

Subject to:

- Discrete choice model
- Batt charging dynamics
- Departure time
- Energy Request



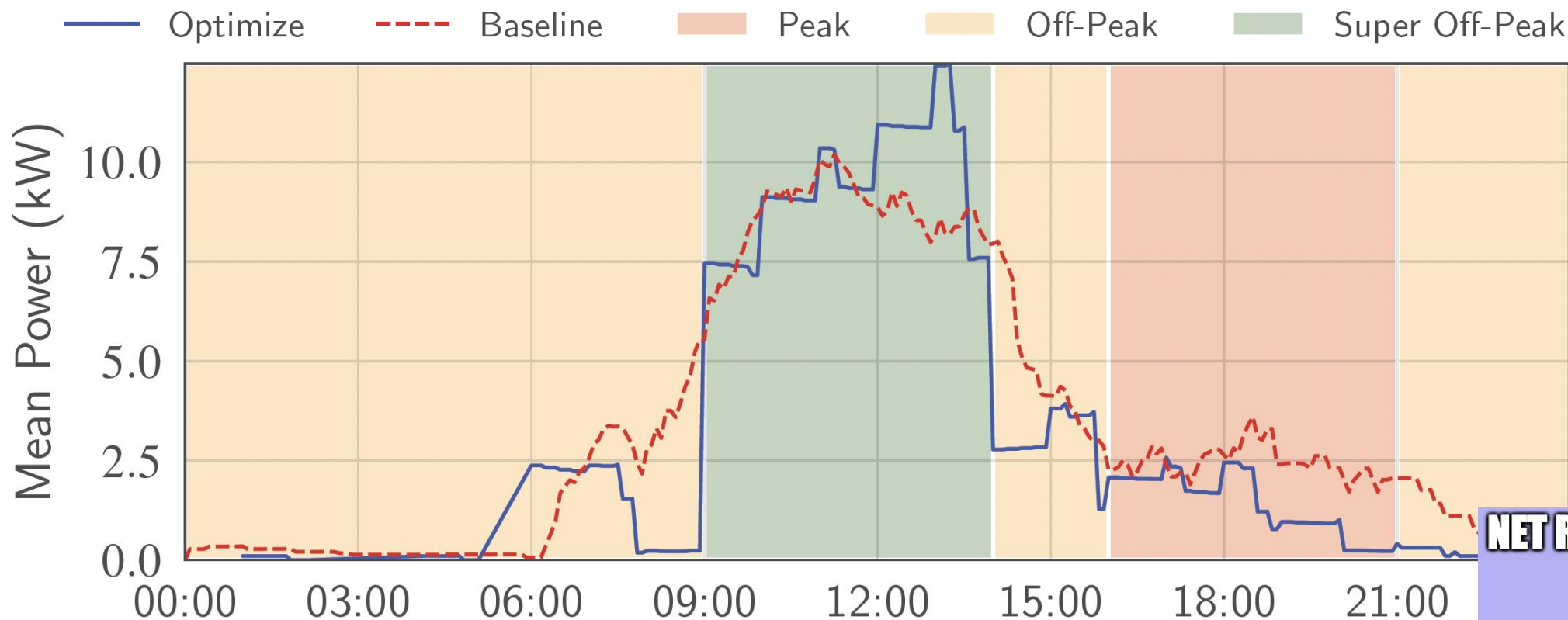
Net Revenue Maximization: Baseline vs. Optimize



Baseline Net Revenue: \$16.32 / day
Optimize Net Revenue: \$24.02 / day

+47% increase in net revenue
+17% increase in gross revenue
-18% decrease in cost

Load Shifting: Baseline vs. Optimize



Fraction of Energy Delivered	Baseline	Optimize
Super Off-peak (09:00 – 14:00)	56%	73%
Off-Peak (14:00 – 16:00, 21:00 – 09:00)	28%	17%
Peak (16:00 – 21:00)	16%	10%

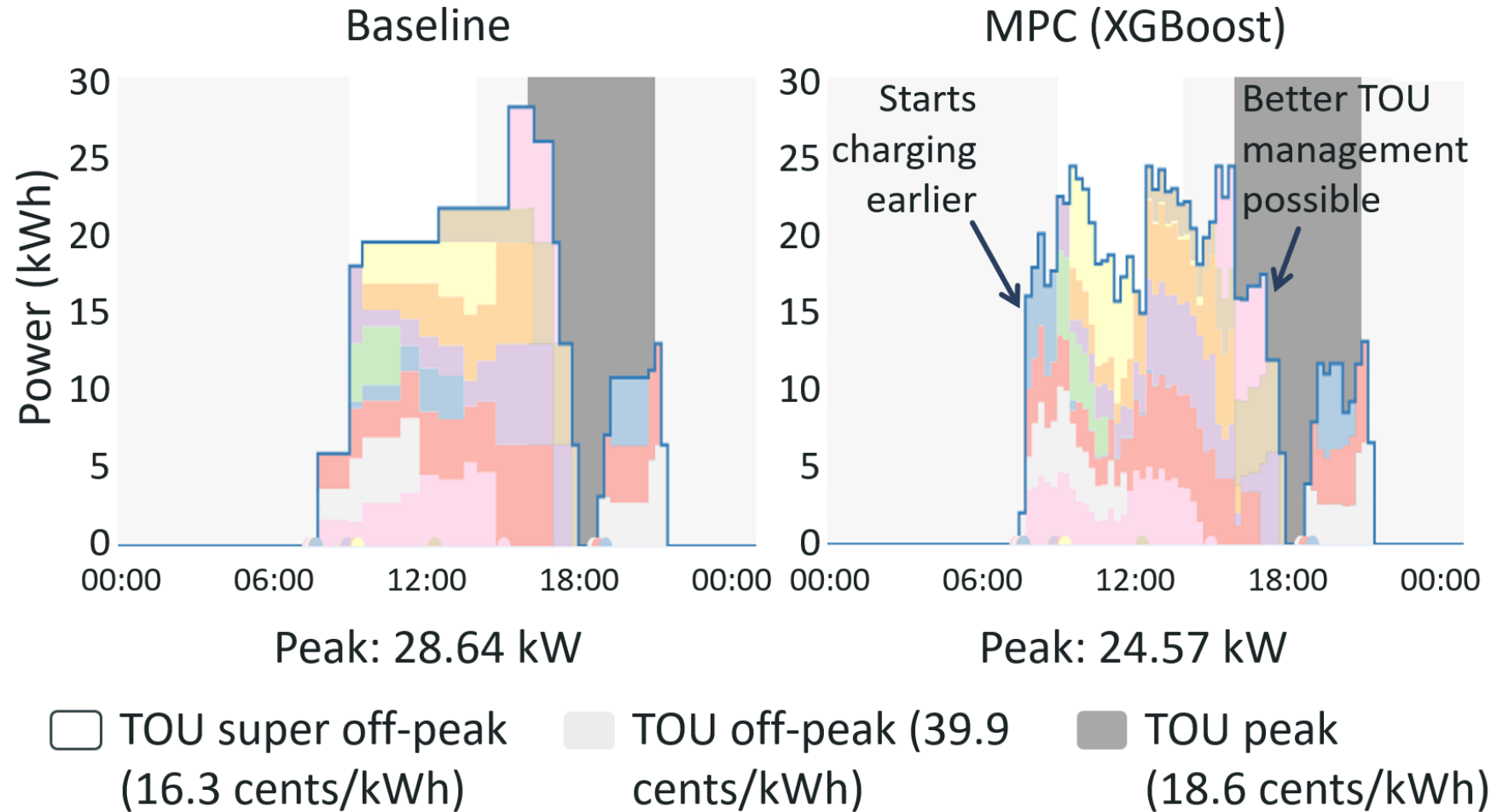
NET REVENUE INCREASED!



**WITH AI/OPTIMIZATION-BASED
PRICE & POWER**

imgflip.com

Optimizing Price & Power can reduce peak power!



Peak Reduction via Price AND Power Optimization

Control Scheme	Mean Cost/Revenue (\$)					Change from Baseline (%)		
	Demand Charge	TOU	Revenue	Cost	Profit	Demand Charge	TOU	Cost
Baseline	624	1,984	4,309	2,608	1,702	0.00	0.00	0.00
Threshold	628	1,987	4,331	2,615	1,716	-0.71	0.16	+0.29
Softplus	609	2,017	4,446	2,626	1,819	-2.34	1.67	+0.71
MPC (Naive)	519	1,970	4,253	2,488	1,765	-16.90	-0.71	-4.58
MPC (Linear)	521	1,970	4,362	2,491	1,870	-16.50	-0.68	-4.47
MPC (XGBoost)	533	1,956	4,434	2,489	1,944	-14.57	-1.39	-4.55

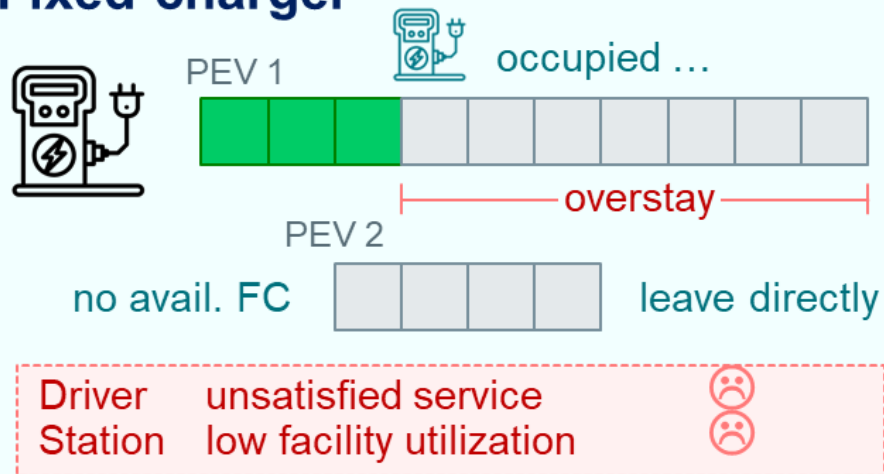
Innovative Ideas to Enable Demand Flexibility

Mobile & Robotic chargers to alleviate overstay

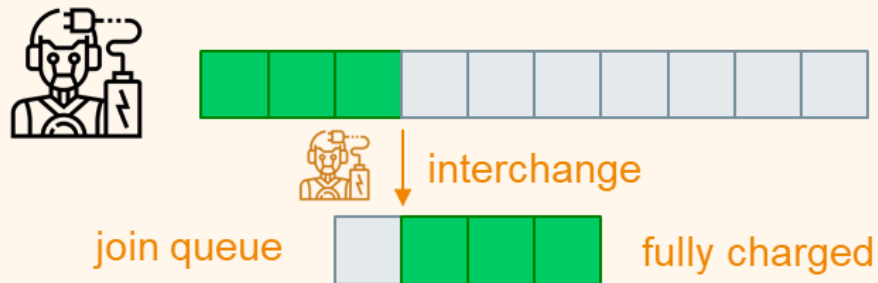


Optimal Operation and Planning of a Robotic Charging System to Alleviate Overstay

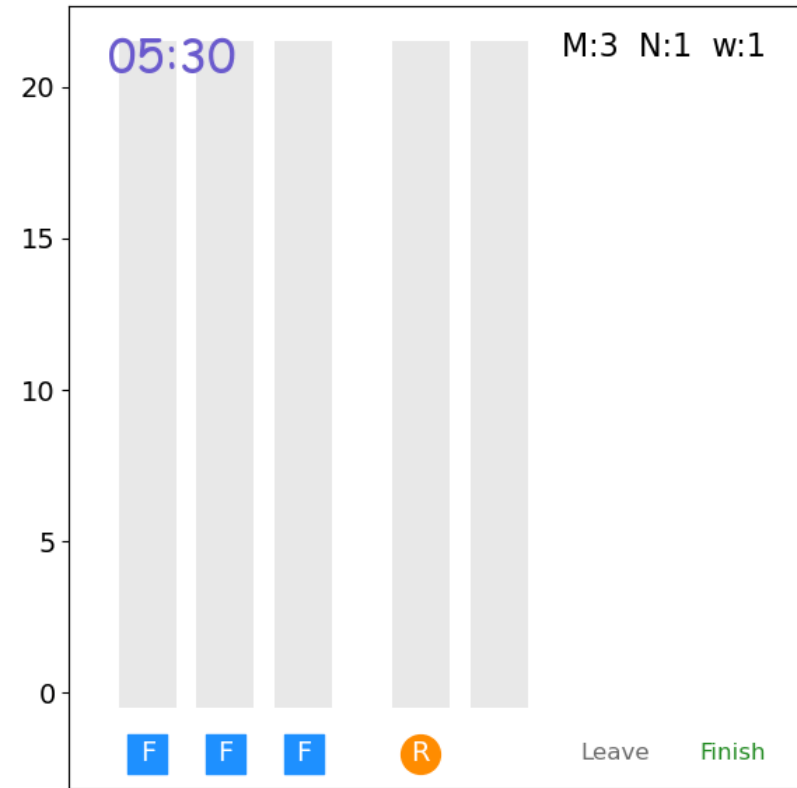
Fixed-charger



Robo-charger



- [Optimal scheduling](#), incl. charger assignment, plug-in schedules, & power profile optimization of Robo-chargers & Fixed-chargers
- [Optimal combo. of Robo- & Fixed-chargers](#) achieve minimum total cost of ownership (TCO)



Mobile & Robotic chargers to alleviate overstay (real-world examples)



Source: [Inductive Robotics](#)



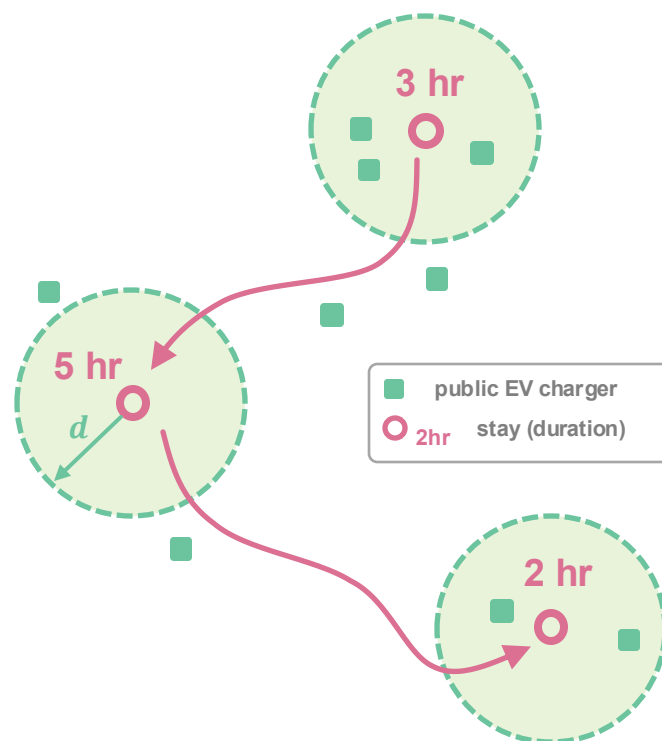
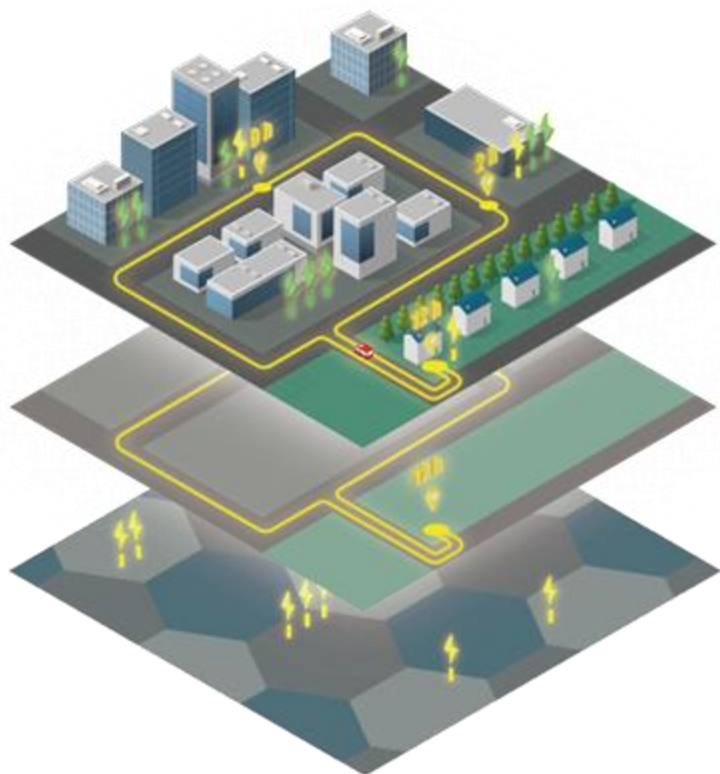
Source: [ElectricFish](#)

Note: Robo-chargers open the opportunity for demand management
(all parked EVs can be (dis)charged)

Disclaimer: we did not involve in any of above products / businesses. All resources are from the Internet.

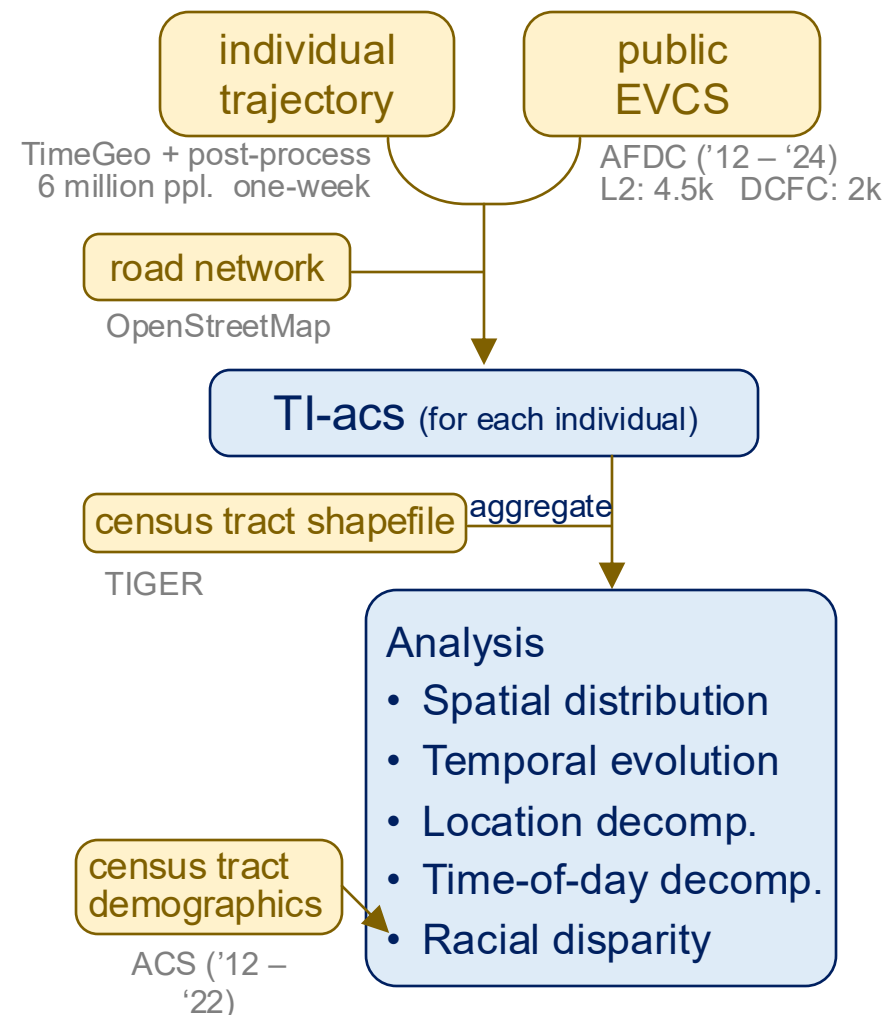
Trajectory-Integrated accessibility

-- Methodology Overview



$$\text{TI-acs [hours]} = 3 \times 1 + 5 \times 0 + 2 \times 1 = 5$$

$$\text{TI-acs [ports]} = \frac{1}{3+5+2} (3 \times 3 + 5 \times 0 + 2 \times 2) = 1.3$$



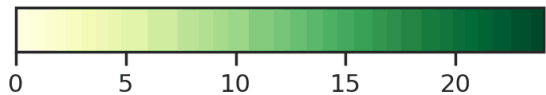
Mobility-fulfillment **market:** public charger accessibility improved over time

2012 (L2)



CT avg. Accessible Hours
within 1000 m range

mean: 1.25 [25, med, 75] = [0.30, 0.55, 1.07]

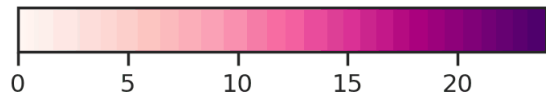


2012 (DCFC)



CT avg. Accessible Hours
within 1000 m range

mean: 0.08 [25, med, 75] = [0.02, 0.03, 0.07]



As of **Aug. 2024**, on average,
SF Bay area residents have

L2: 7.5 hrs / day

DCFC: 5.2 hrs / day

when there is at least 1 public
charger within 1 km of their
stays

five years ago (2019):

L2: 4.7 hrs / day

DCFC: 2.2 hrs / day

Motivation: Battery Swapping Opportunity

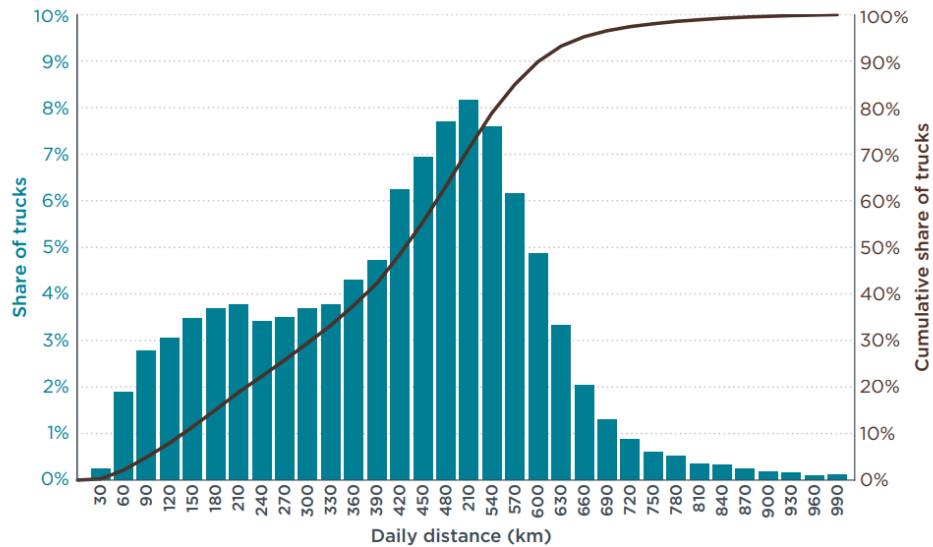
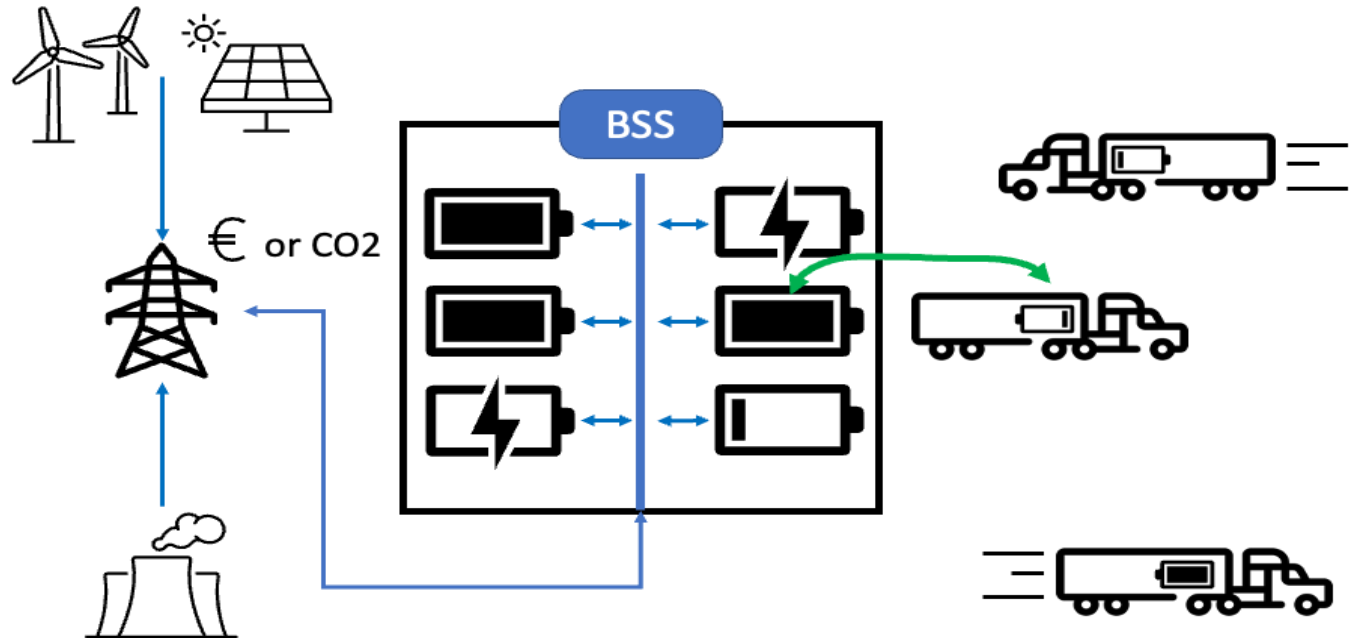


Figure 1. Average truck daily distance from a representative fleet, adapted from Wentzel (2020).

- 90% of trucks drive a daily distance shorter than 600 km / 373 mi (Basma et al., 2021)

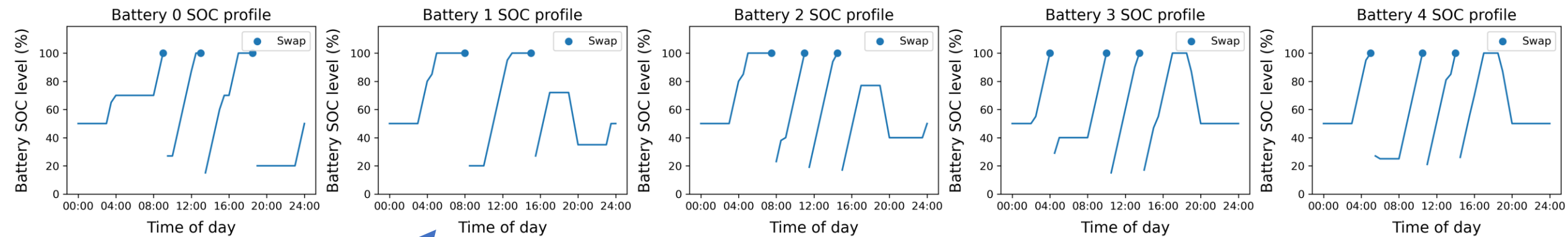


Battery range ~ 300 km / 186 mi
With one swap ~ 600 km / 373 mi

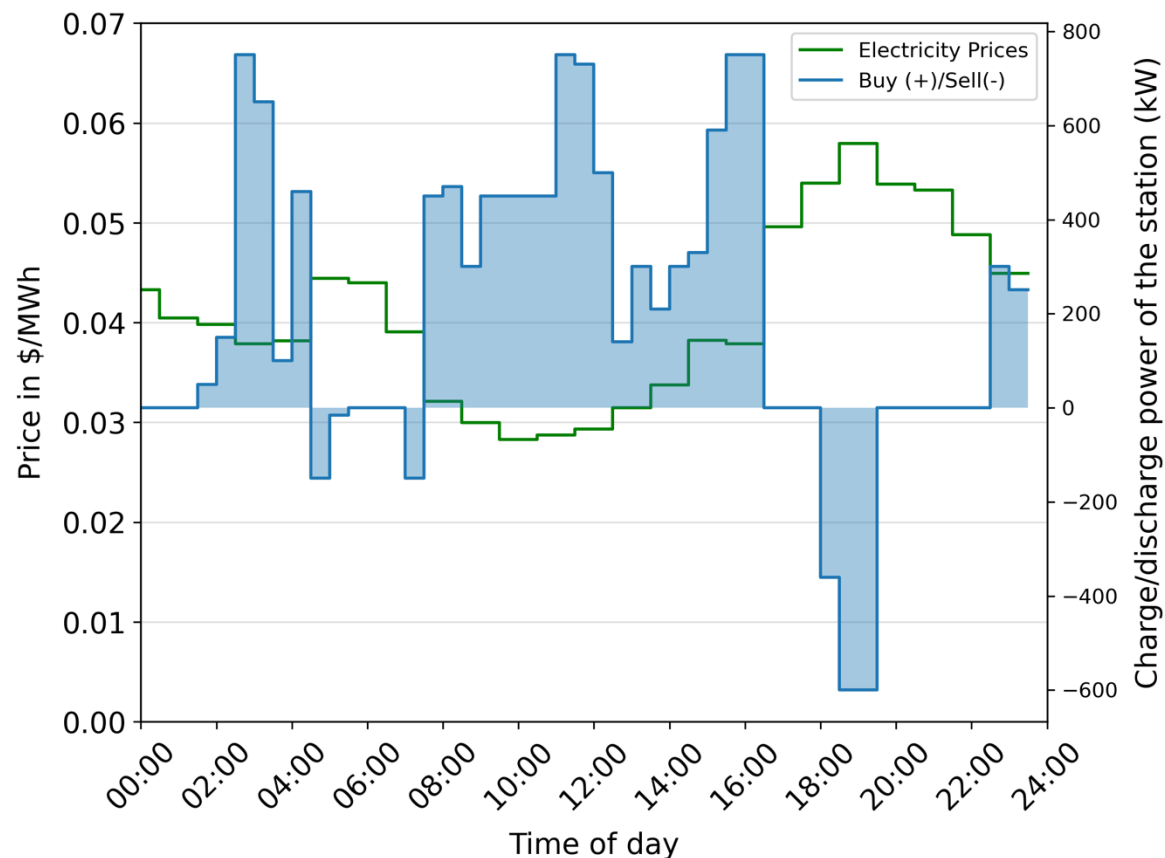
Volvo FM
electric,
540kWh of
battery for
300km –
Solutrans 2021

Stack Services

1. Sell e-mobility service, as a fueling station
2. Sell grid services, as a battery storage system



- Charged batts are swapped out for discharged batts
- Discharge around 18h00



- Station purchases power at low prices, sells at high prices

Outline

1. Introduction
2. The Duck Curve Problem
3. Power Grid Impacts of EVs
4. Saving the Grid with SlrpEV
5. Summary

Summary

Will EVs Break the Power Grid?

- Transportation contributes ~45% of CA GHG emissions
- Trajectory towards 100% ZEVs will require significant power grid infrastructure upgrades
- Queue of clean energy is huge, but interconnection is slow
- Massive need and opportunity to create demand flexibility from EV Charging
 - SlrpEV, robo/mobile chargers, battery swapping, behavioral nudging along mobility trajectories



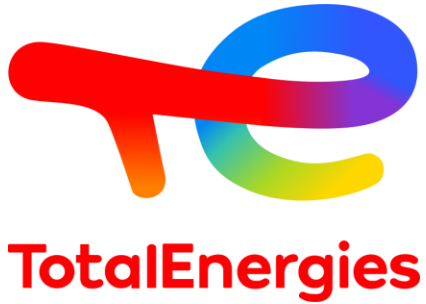
Collaborate with us to...

- Reduce transportation CO2e via electrification
- Creative innovations in e-mobility





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Disclaimer

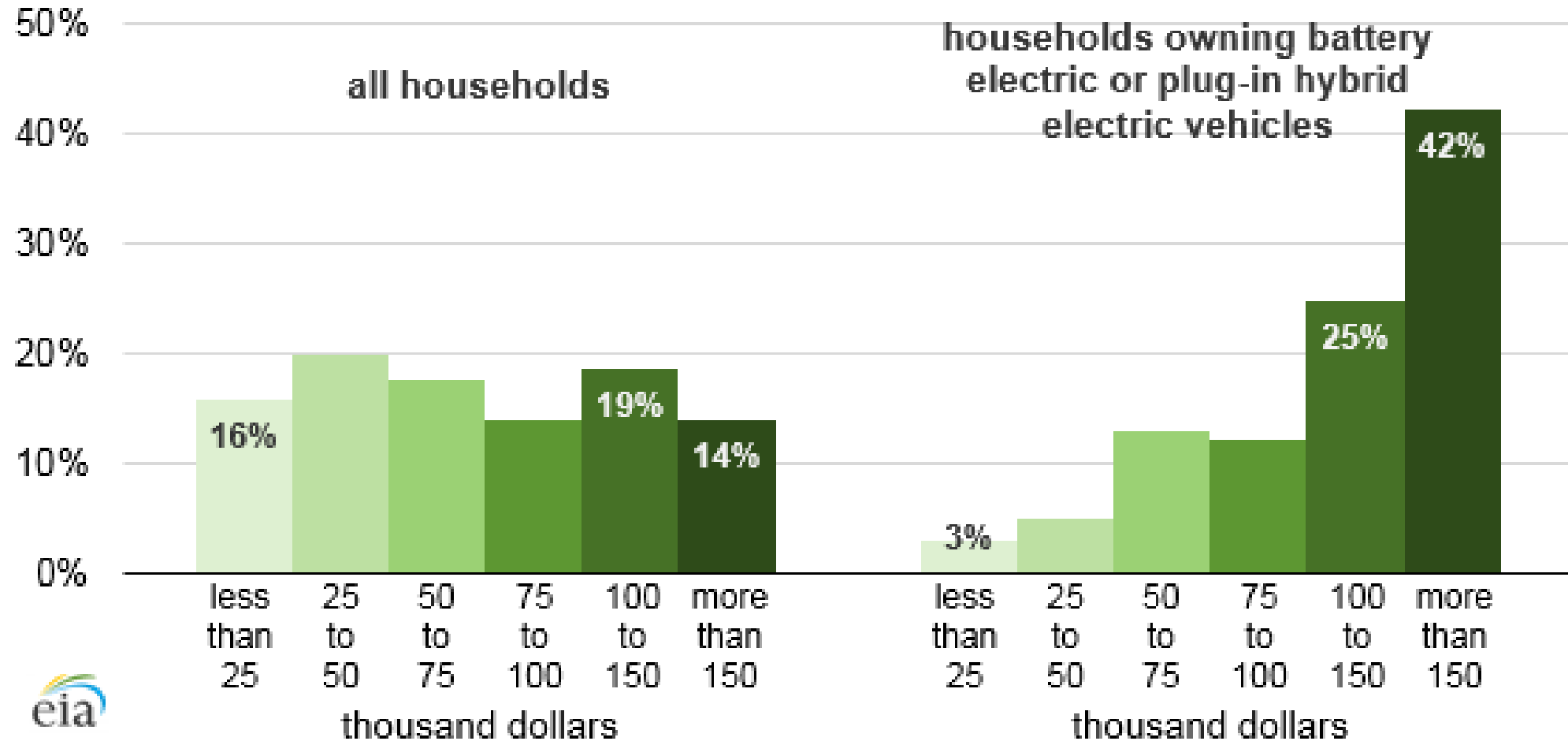
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Access to ZEVs by **INCOME**

U.S. household income distribution, 2017

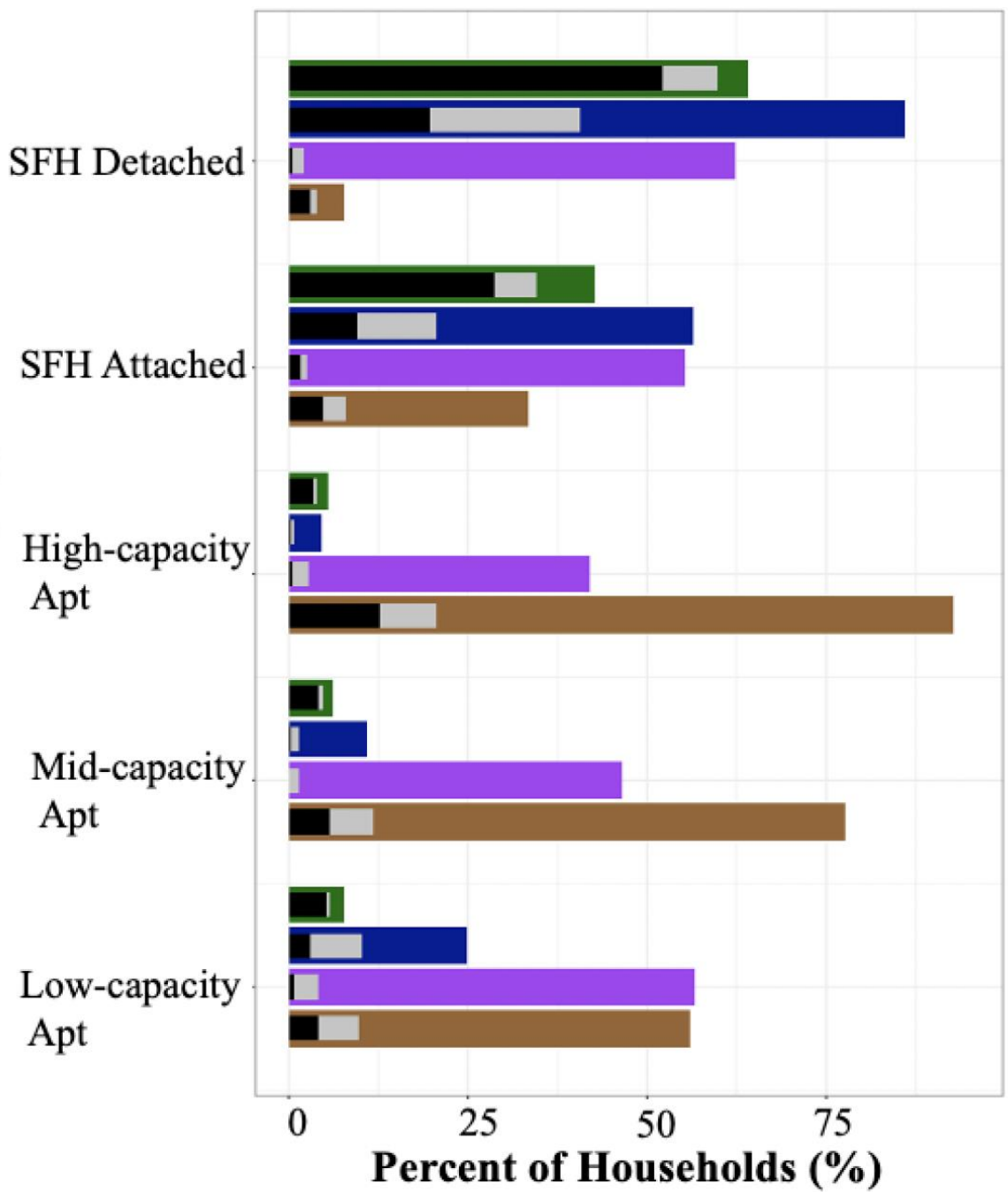


Equitable Access to ZEVs by HOUSING

At home, do you
have an electrical
plug near your car?

Apartment dwellers
largely LACK
electrical access for
charging

Housing Type



Legend

- Personal Garage
- Driveway/Carport
- On-Street
- Parking Garage/Lot

Share w/
Available
Electrical
Access

Share w/
Potential
Electrical
Access

Impact of electric vehicle charging demand on power distribution grid congestion

Y. Li, A. Jenn, "Impact of electric vehicle charging demand on power distribution grid congestion," *Proceedings of the National Academy of Sciences*, April 2024. DOI: [10.1073/pnas.2317599121](https://doi.org/10.1073/pnas.2317599121). Edited by SJM.



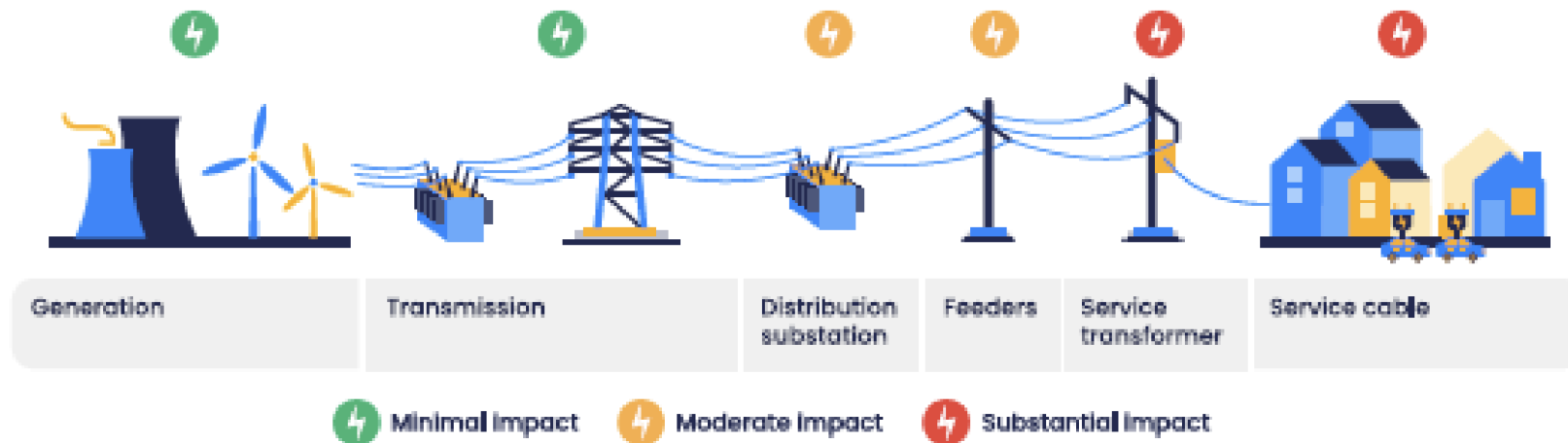
Yanning Li



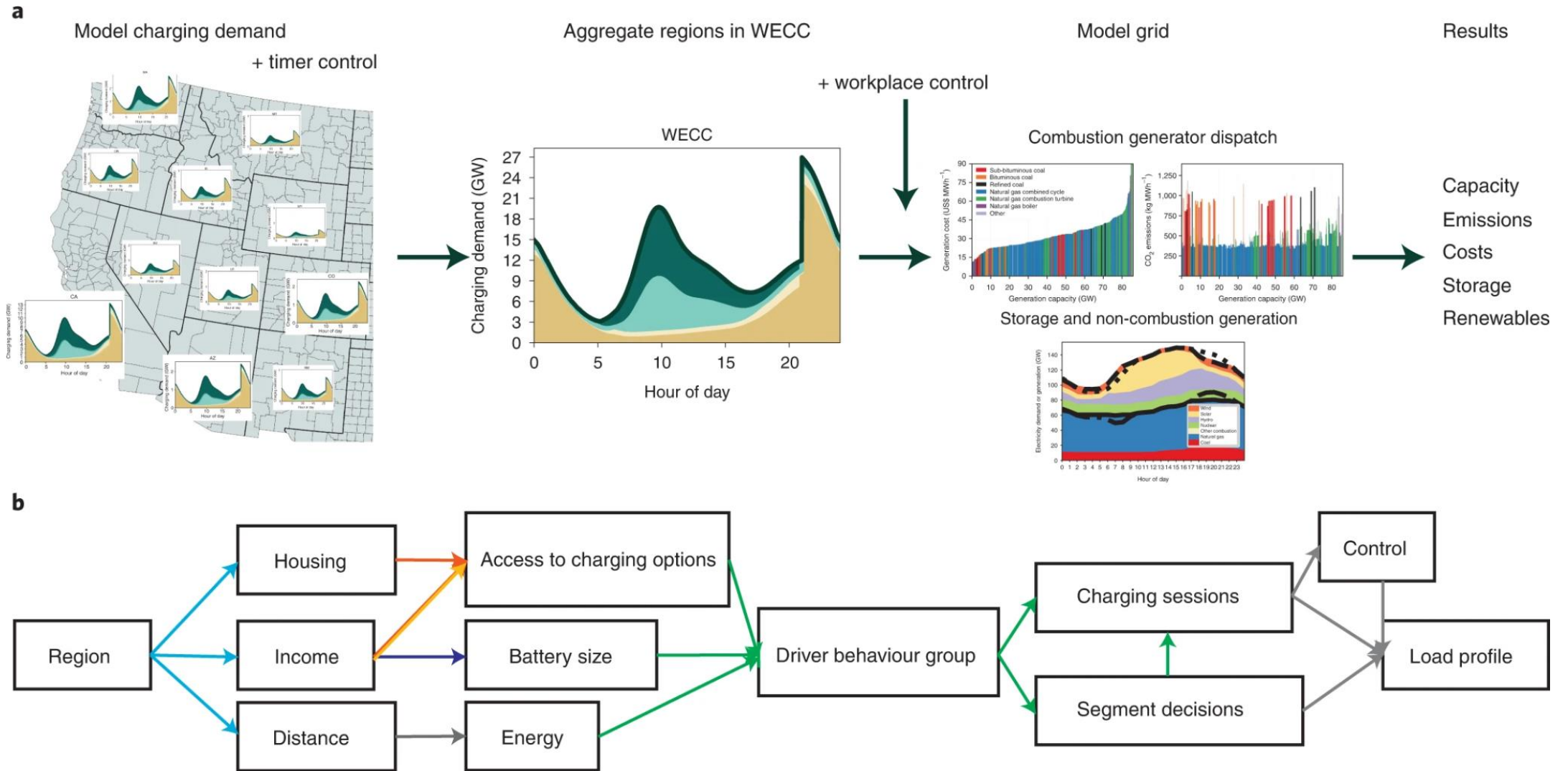
Alan Jenn



Edited by Scott Moura



How does Large-Scale EV Adoption Impact Peak Demand



Source: S. Powell, G. V. Cezar, L. Min, et al. *Charging infrastructure access and operation to reduce the grid impacts of deep electric vehicle adoption*. Nature Energy (2022). <https://doi.org/10.1038/s41560-022-01105-7>