

Adaptive Charging Network

Steven Low

CMS, EE



Caltech

January 2023



Acknowledgement



G. Lee



C. Jin



T. Lee



D. Lee



Acknowledgement



G. Lee



C. Jin



Z. Lee



T. Li



J. Pang



Z. Low, Cornell



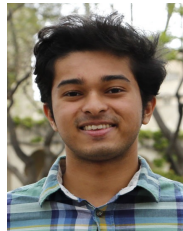
K. Erliksson, Lund



T. Lee



D. Lee



S. Sharma



D. Guo



C. Ortega



D. Johansson, Lund



R. Lee



D. Chang

and many others ...



Agenda

ACN: Caltech testbed

- Testbed to commercial deployment

ACN: Research Portal

- Data, Sim, Live

ACN: pricing demand charge

- Monthly billing at workplaces

Unbalanced 3-phase modeling

- Motivation, 3-phase network models





ACN testbed

IEEE TRANSACTIONS ON SMART GRID, VOL. 12, NO. 5, SEPTEMBER 2021

4339

Adaptive Charging Networks: A Framework for Smart Electric Vehicle Charging

Zachary J. Lee^{ID}, *Graduate Student Member, IEEE*, George Lee, Ted Lee^{ID}, Cheng Jin, Rand Lee, Zhi Low^{ID}, Daniel Chang, Christine Ortega, and Steven H. Low^{ID}, *Fellow, IEEE*

2016 GlobalSIP Conference:

Adaptive Charging Network for Electric Vehicles

George Lee^{1,2}, Ted Lee², Zhi Low³, Steven H. Low², and Christine Ortega²

¹PowerFlex Systems

²Division of Engineering & Applied Science, Caltech

³Math Department, Cornell



ACN Research Portal

2019 ACM e-Energy:

ACN-Data: Analysis and Applications of an Open EV Charging Dataset

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IEEE TRANSACTIONS ON SMART GRID, VOL. 12, NO. 6, NOVEMBER 2021

5113

ACN-Sim: An Open-Source Simulator for Data-Driven Electric Vehicle Charging Research

Zachary J. Lee^{ID}, Sunash Sharma^{ID}, Daniel Johansson, and Steven H. Low^{ID}, *Fellow, IEEE*



ACN Pricing



ELSEVIER

Electric Power Systems Research

Volume 189, December 2020, 106694



Pricing EV charging service with demand charge ☆

Zachary J. Lee ^a  , John Z.F. Pang ^b , Steven H. Low ^{a, b} 

PSCC 2020



Unbalance 3-phase modeling

Power System Analysis

A Mathematical Approach

Steven H. Low

DRAFT available at: <http://netlab.caltech.edu/book/>

Corrections, questions, comments appreciated!



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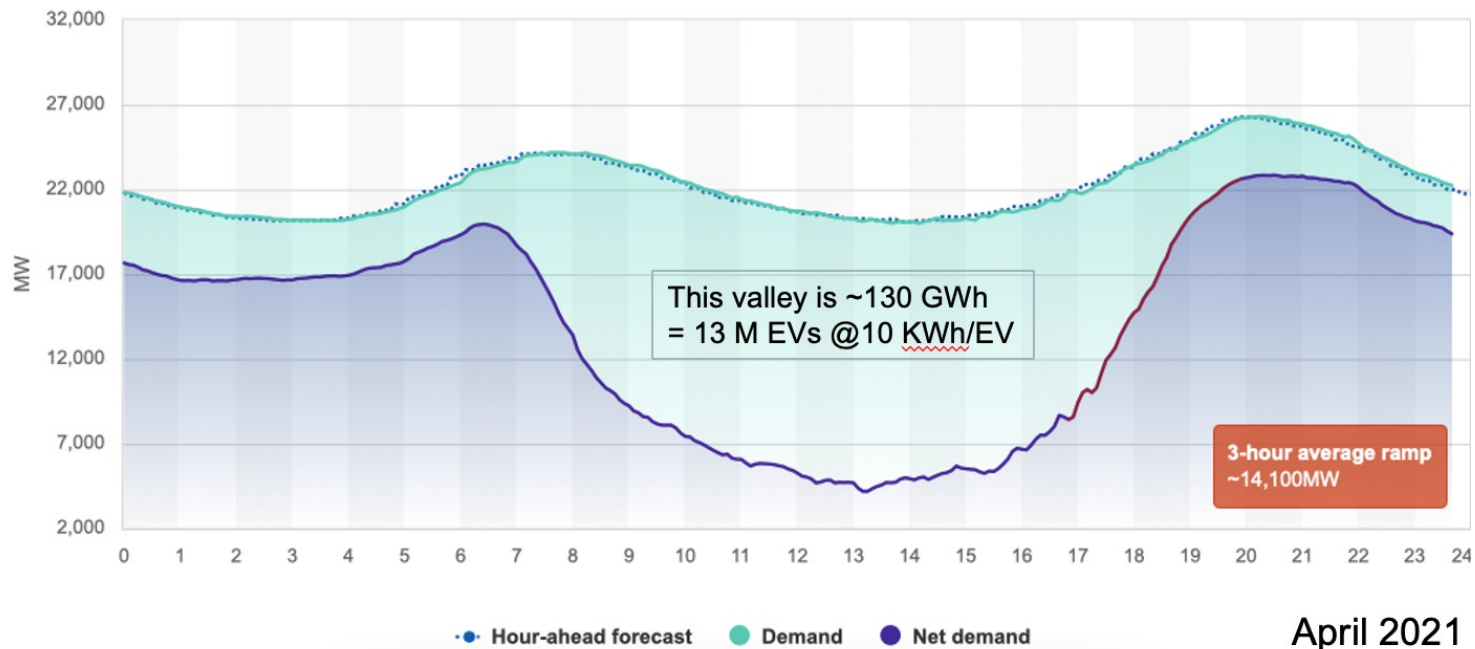




Workplace charging

CA commitment

- ~~50%~~ ^{60%} renewables by 2030, 100% by 2045
- 1.5M ZEV by 2025, 5M by 2030 (CA has ~15M cars)

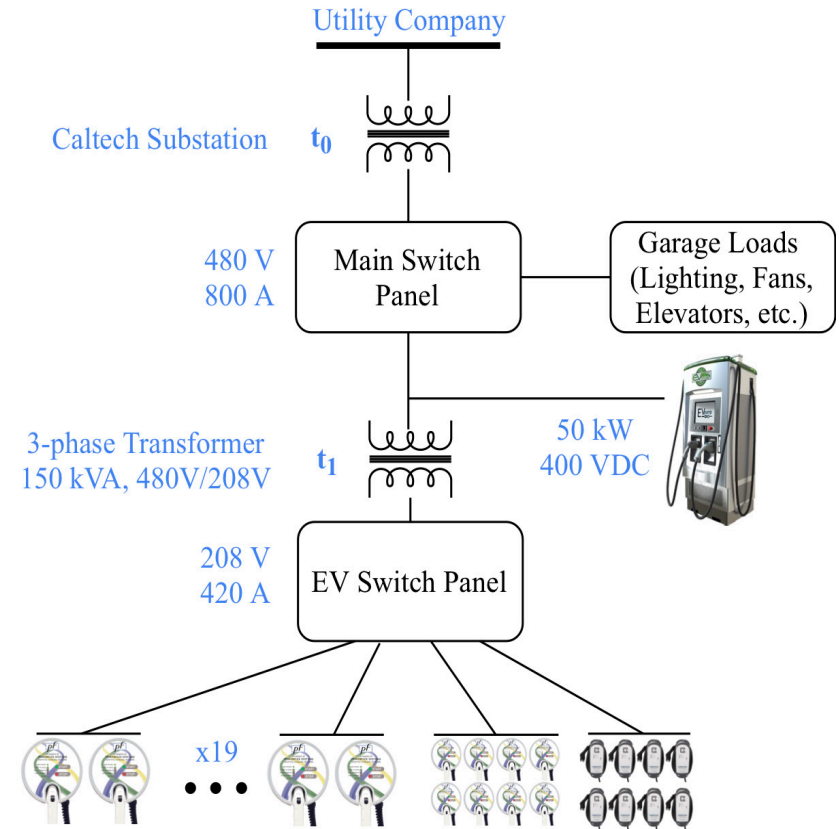


Drivers twice as likely to get EV when workplace charging is available

(EDF Renewables survey Feb 2018)



Caltech ACN: physical system





Caltech ACN: cyber system

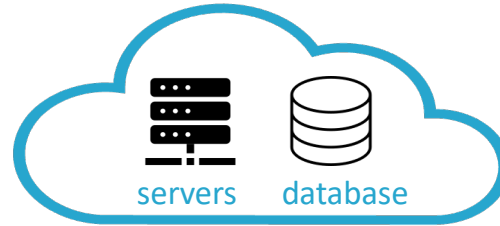
Model predictive control: QCQP

$$\begin{aligned} & \min_{r \geq 0} C(r) \\ \text{subject to} & \quad r_i(t) \leq \bar{r}_i(t) \quad \forall i, \forall t \\ & \quad \sum_t r_i(t) \delta = e_i \quad \forall i \\ & \quad \sum_i r_i(t) \leq P(t) \quad \forall t \end{aligned}$$

Highly customizable QCQP

- objectives: cost, PV, asap, regularization
- constraints: energy, deadlines, capacities
- determine charging rates for all EVs

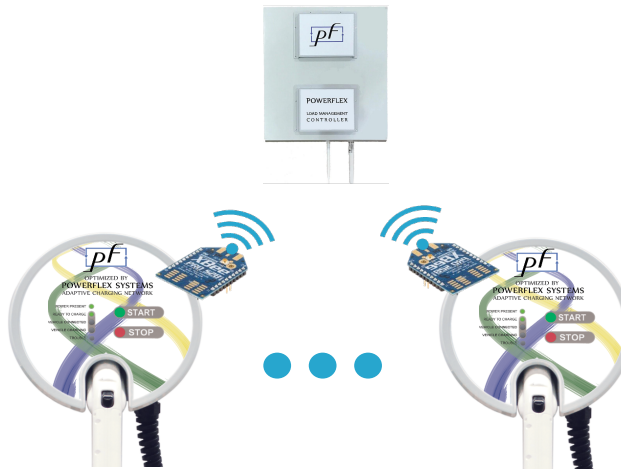
PF cloud



IP/cellular



Garage



Mobile app

First deployment Feb 19, 2016

Online optimization of electric vehicle charging

- Enables mass deployment at lower capital & operating costs
- First pilot @Caltech: 54 adaptive programmable chargers
- 2x 150kVA transformers, breakers, grid sensors, etc

debugging



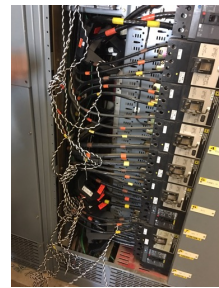
charger



transformer & subpanels



main panel



2020



Figure 1. Photos of the N_Wilson_Garage_01 ACN, which is one of the charging sites used to collect data.

The ACN [Research Portal](#) has three parts:

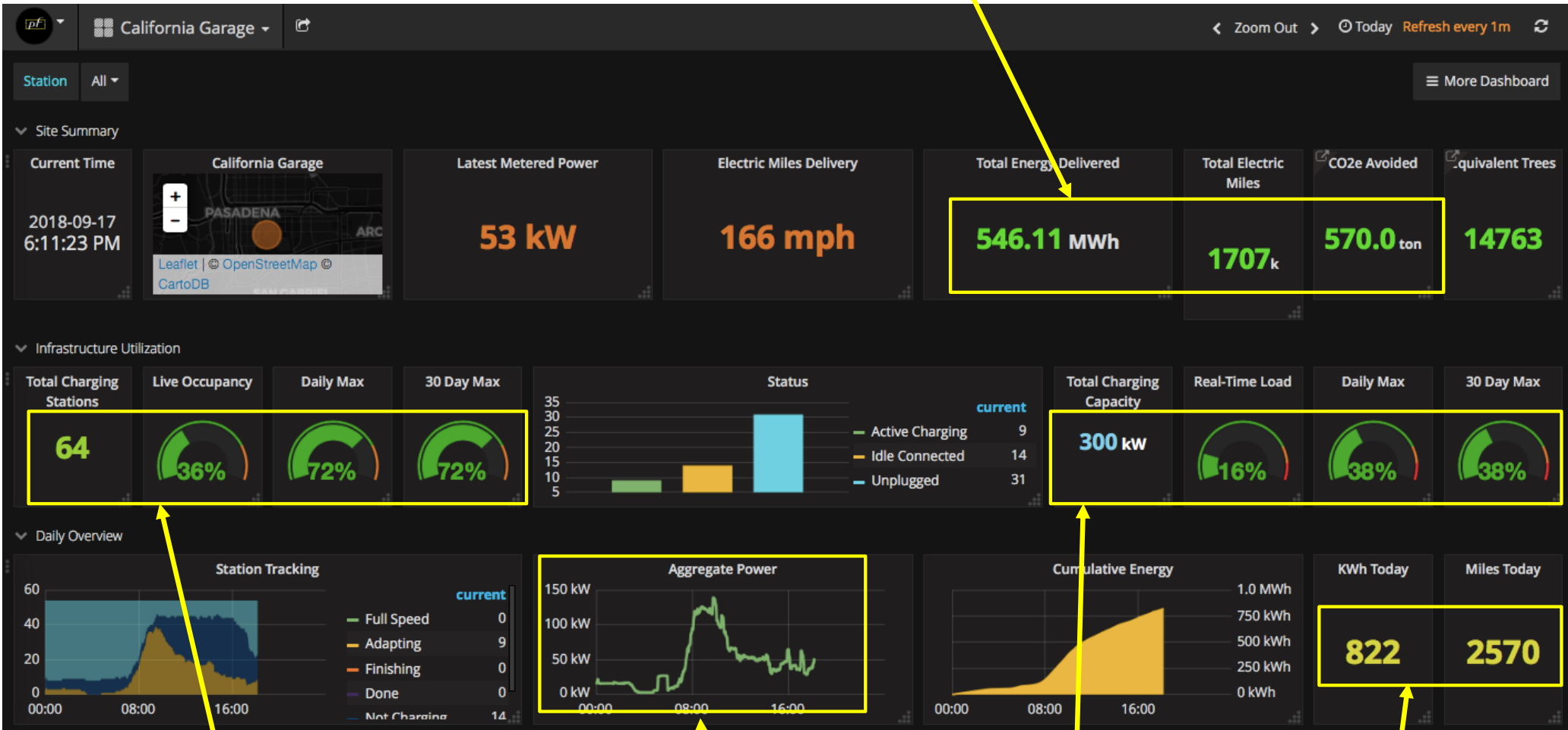
- (1) ACN-Data: a dataset of over 80,000 EV charging sessions (March 2021)
- (2) ACN-Sim: an open-source, data-driven simulation environment
- (3) ACN-Live: a framework for field testing algorithms on physical hardware

March 2021: ACN includes a total of 207 level-2 EVSEs and six DC Fast Chargers (DCFC), and covers seven sites at Caltech, NASA's Jet Propulsion Laboratory, a LIGO research facility, and an office building in Northern California.



Caltech ACN

energy delivered & impact to date



charging station utilization

peak power

power utilization

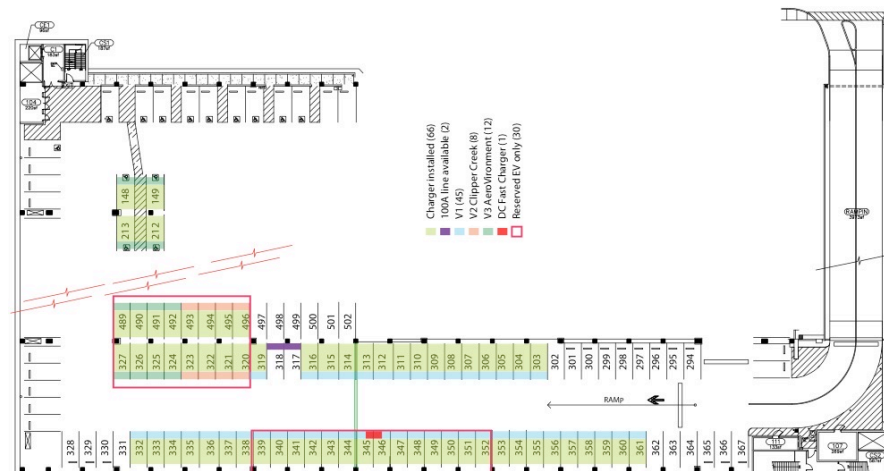
today's energy delivered



Caltech ACN

Spatial utilization snapshot (June 1 – August 31, 2018)

	total	per day	per space	remark
#parking spaces	53			
#days (June 1 – Aug 31, 2018)	92			inc. weekends
#charging sessions	6,103	66	115	> 1 session /space/day
occupancy (space-day)	3,374	37	64	69% occupancy
energy delivered (kWh)	54,562	593	1,029	11 kWh /space/day
#hours occupied	28,407	309	536	5.8 hours /space/day





Caltech ACN



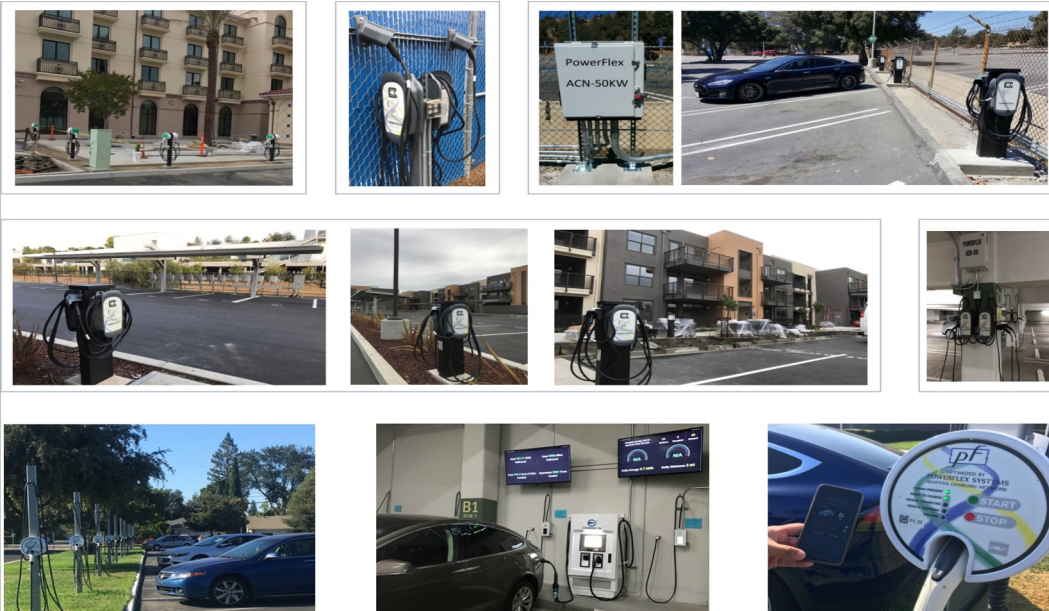
- CA Garage operational since 2016
- Delivered 1 GWh (by July 2020, CA)
- Equivalent to 3.2M miles, 1,000 tons of avoided CO₂e



Caltech ACN



- CA Garage operational since 2016
- Delivered 1 GWh (by July 2020, CA)
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Feb 2020

2,000+
EV CHARGING
STATIONS DEPLOYED

10,000,000+
ELECTRIC MILES
DELIVERED SAFELY

(US wide)



NREL, Golden CO



120 EVSEs



Bay Area high schools



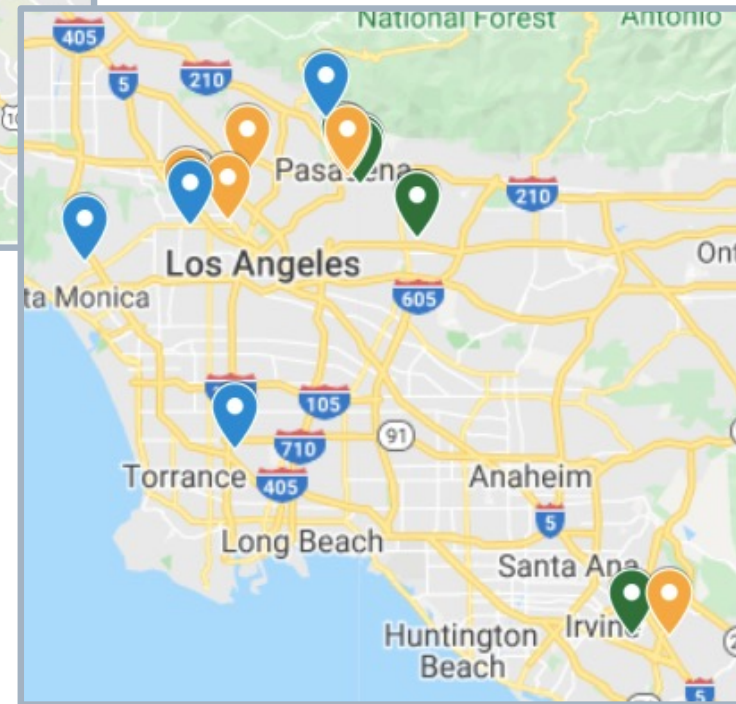
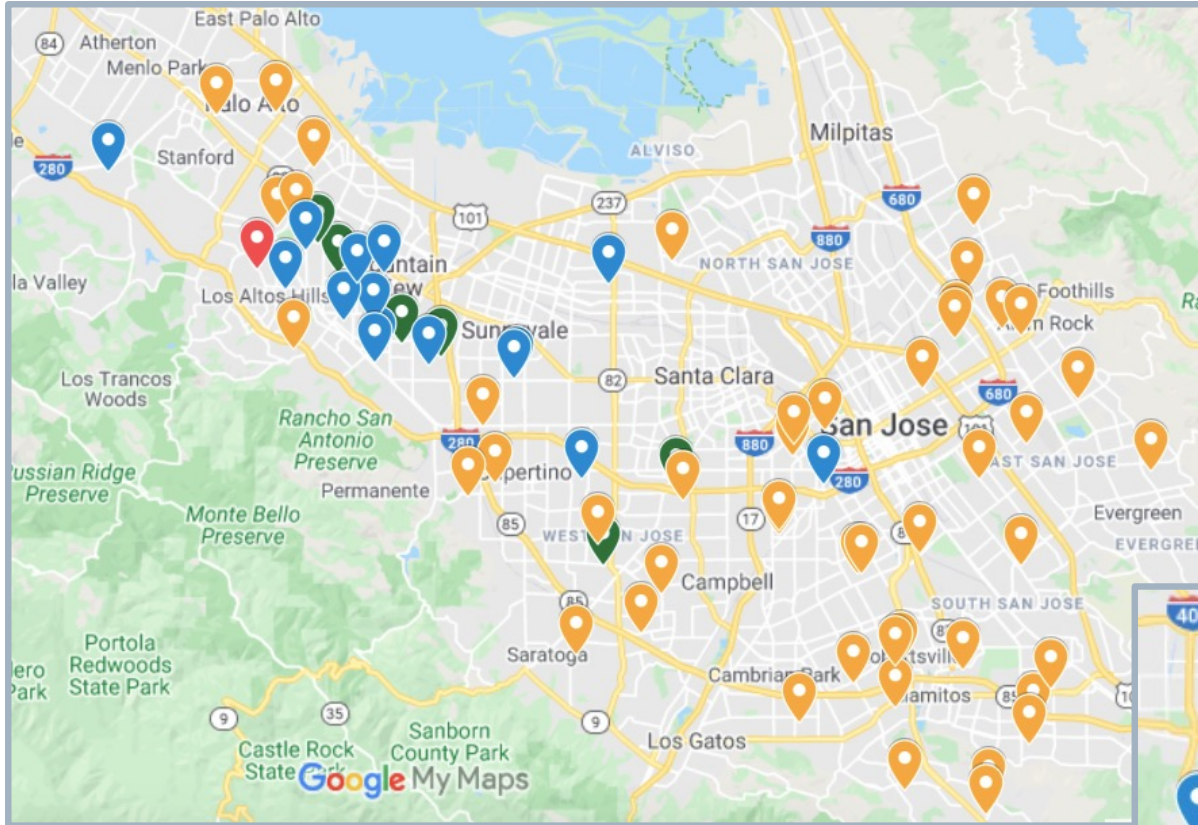
DCFC



Onsite PV



Deployment in CA

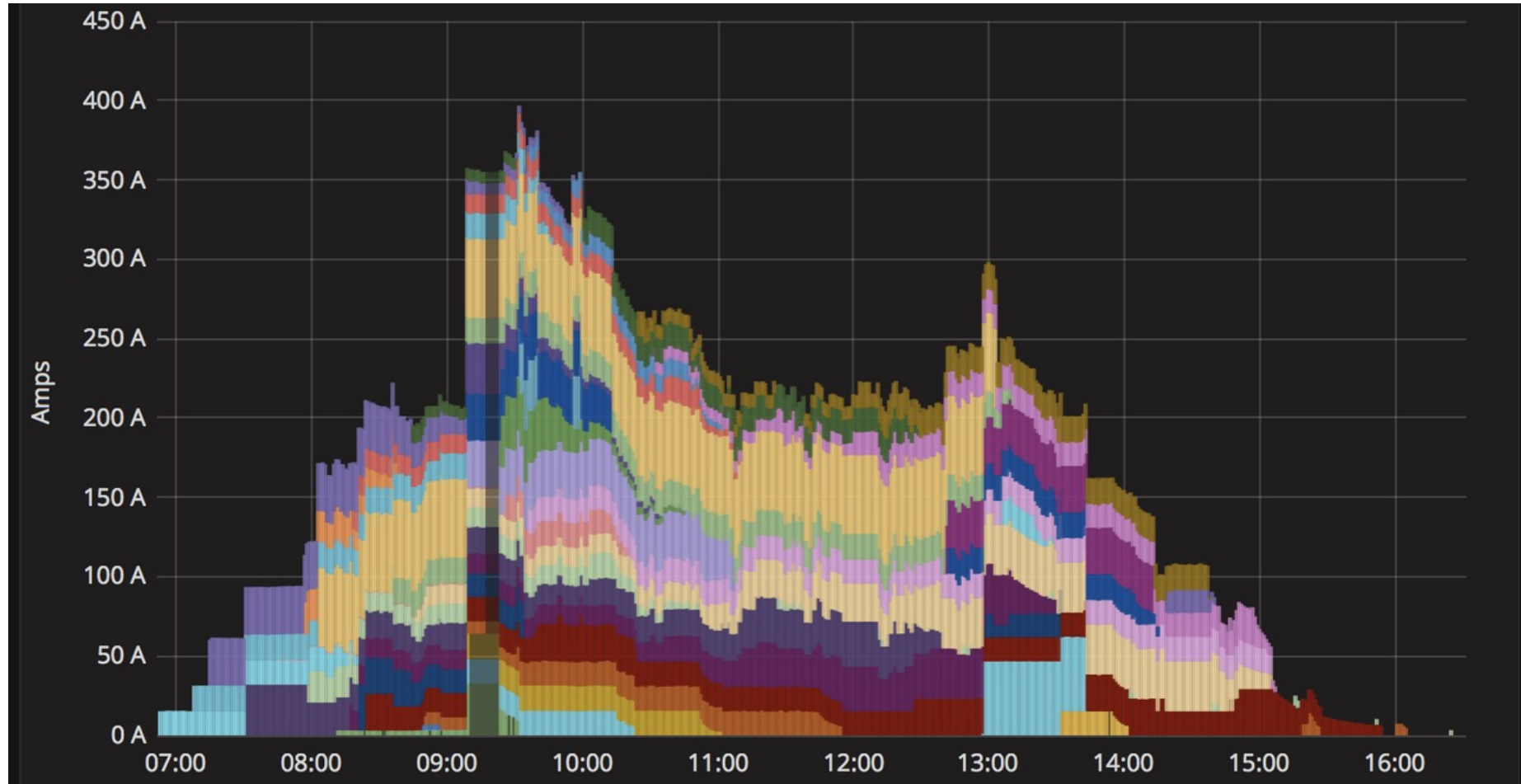


 powerflex
EDF renewables deployment, Sept 2018



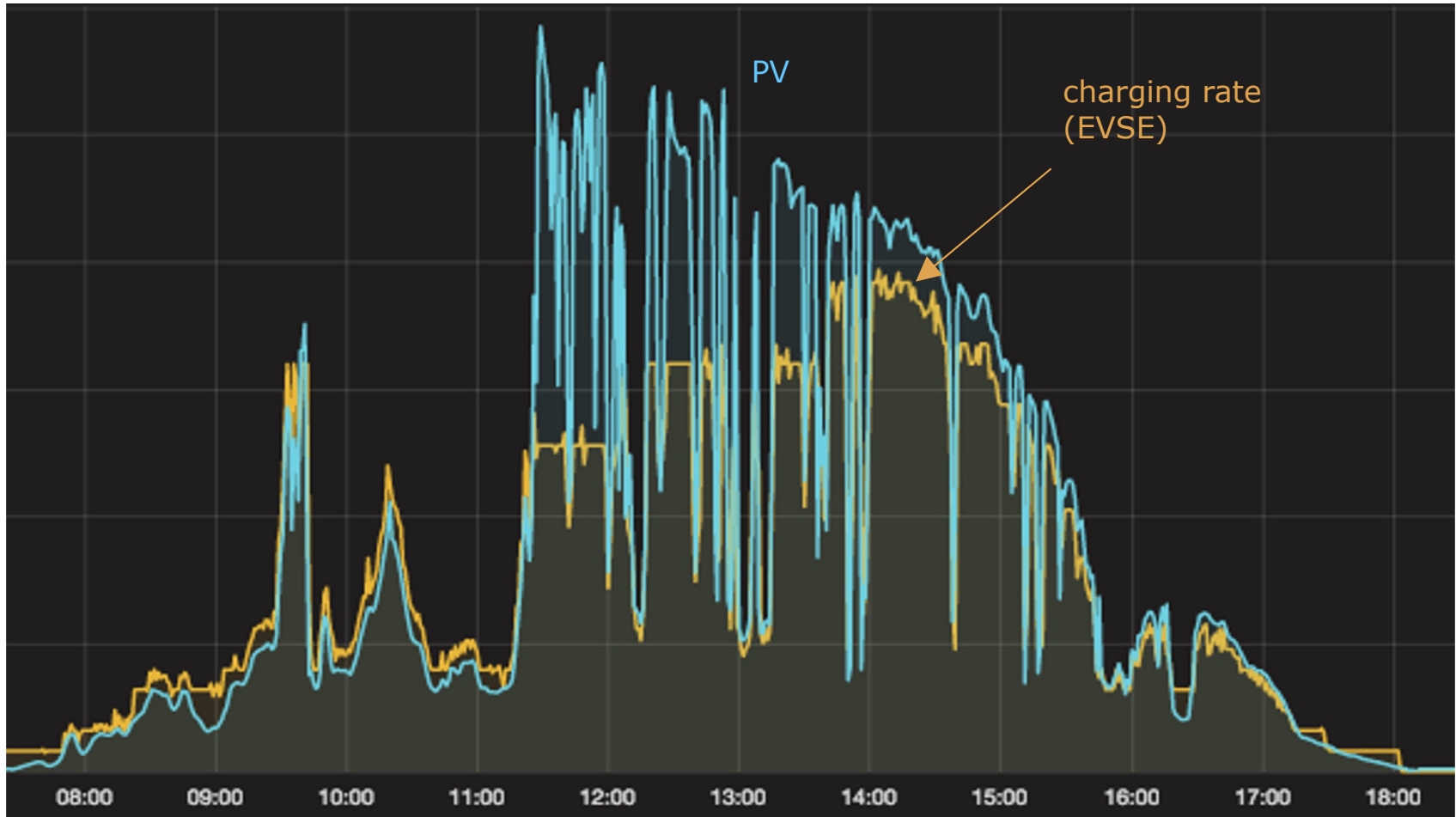


Adaptive charging





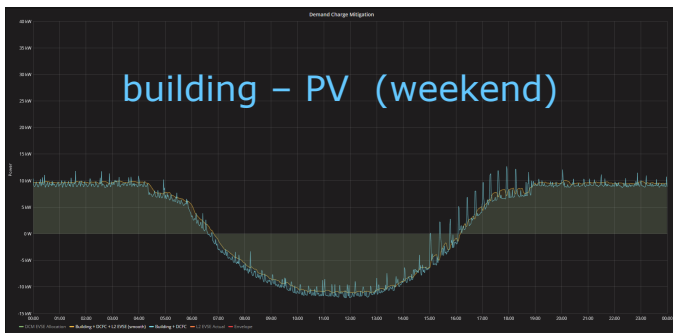
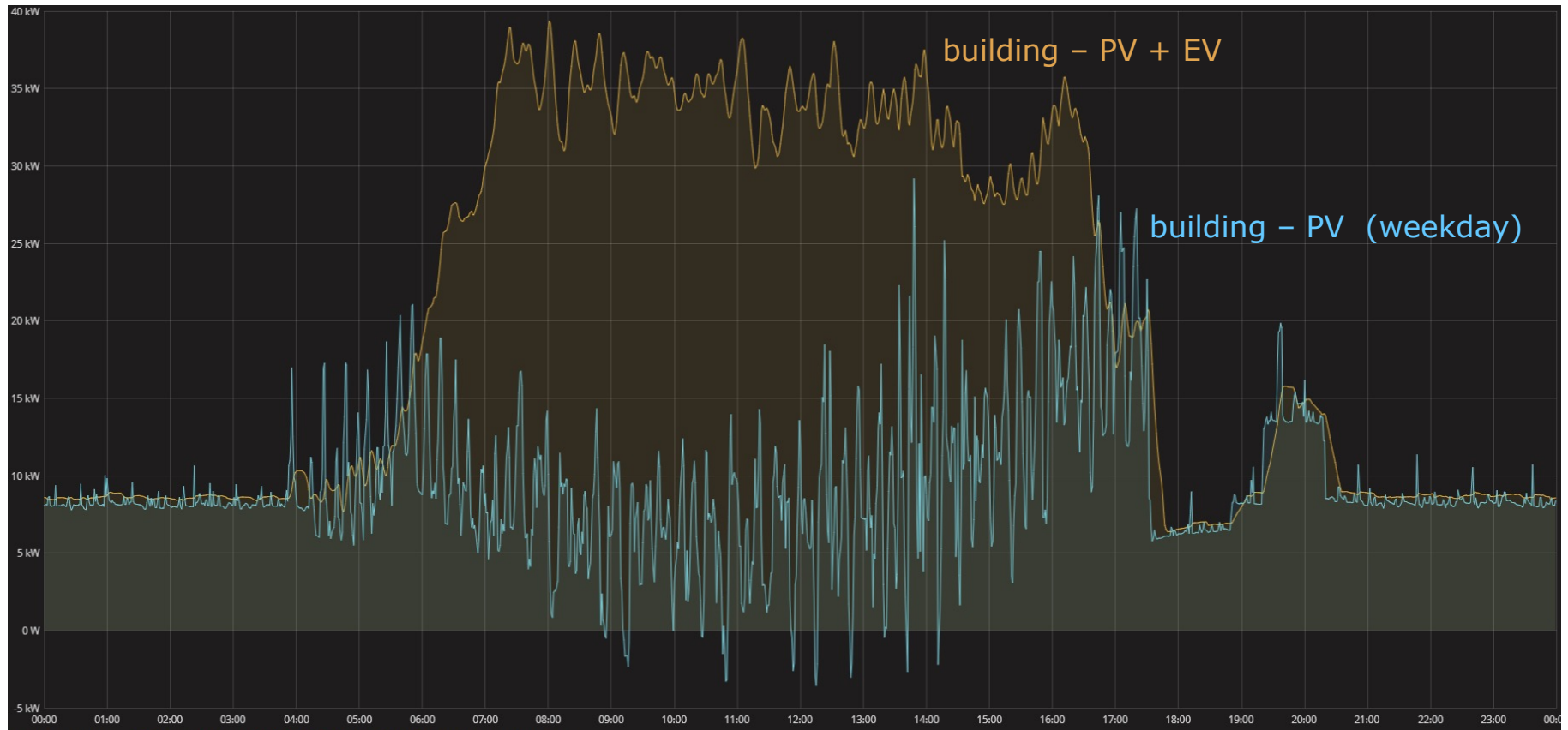
Online tracking



Real-time tracking of PV generation at JPL (10/2016)



Duck Curve & DCM

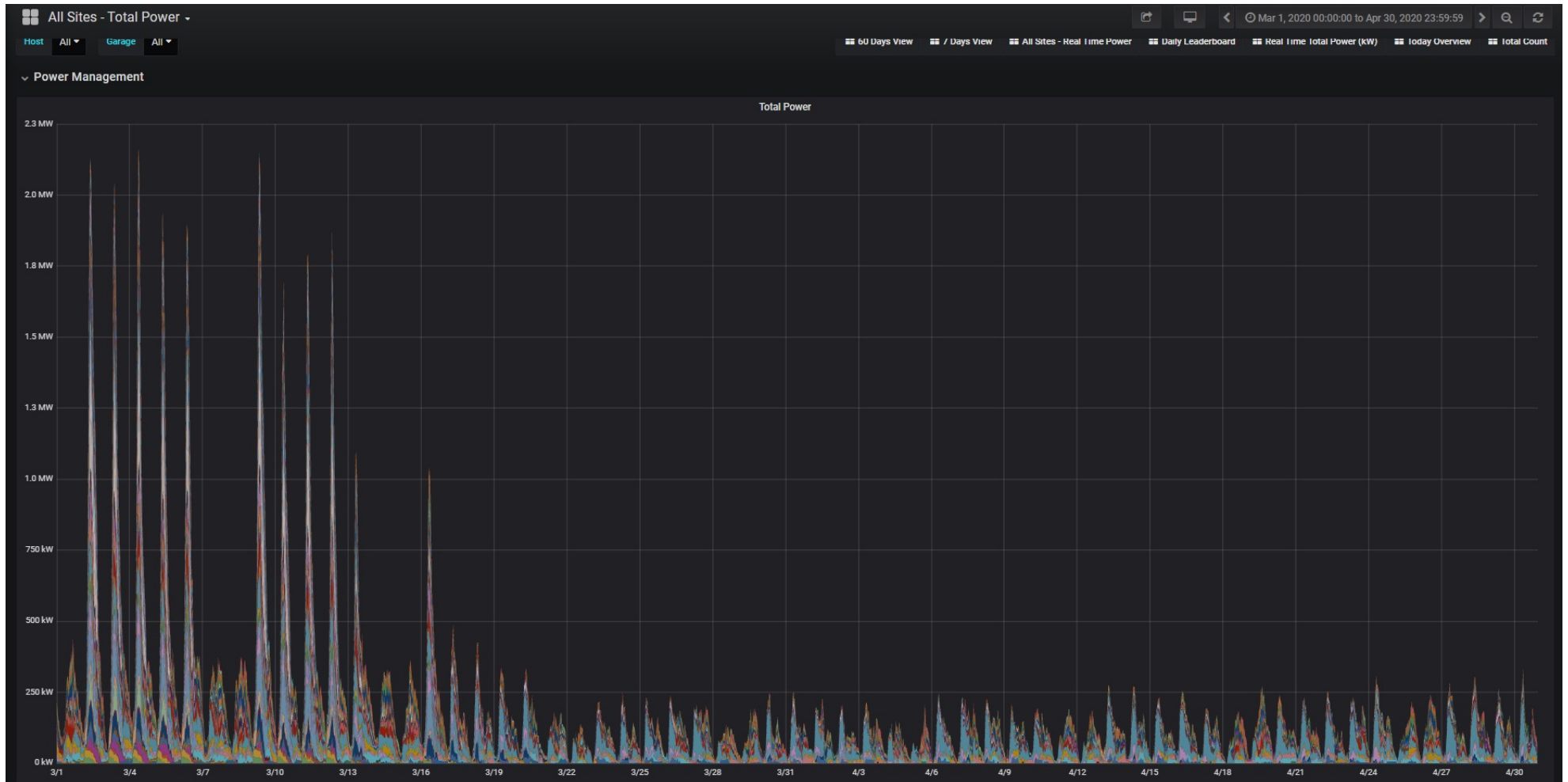


Weekend Duck Curve: building load (10kW) – PV

NREL: demand charge mitigation (Nov 2018)

- Fill Duck Curve valley and maintain net load between 30 kW – 40 kW
- On weekdays: building load is much higher and much more volatile

COVID hit

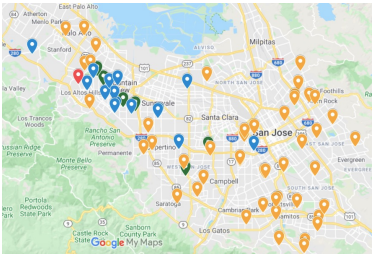
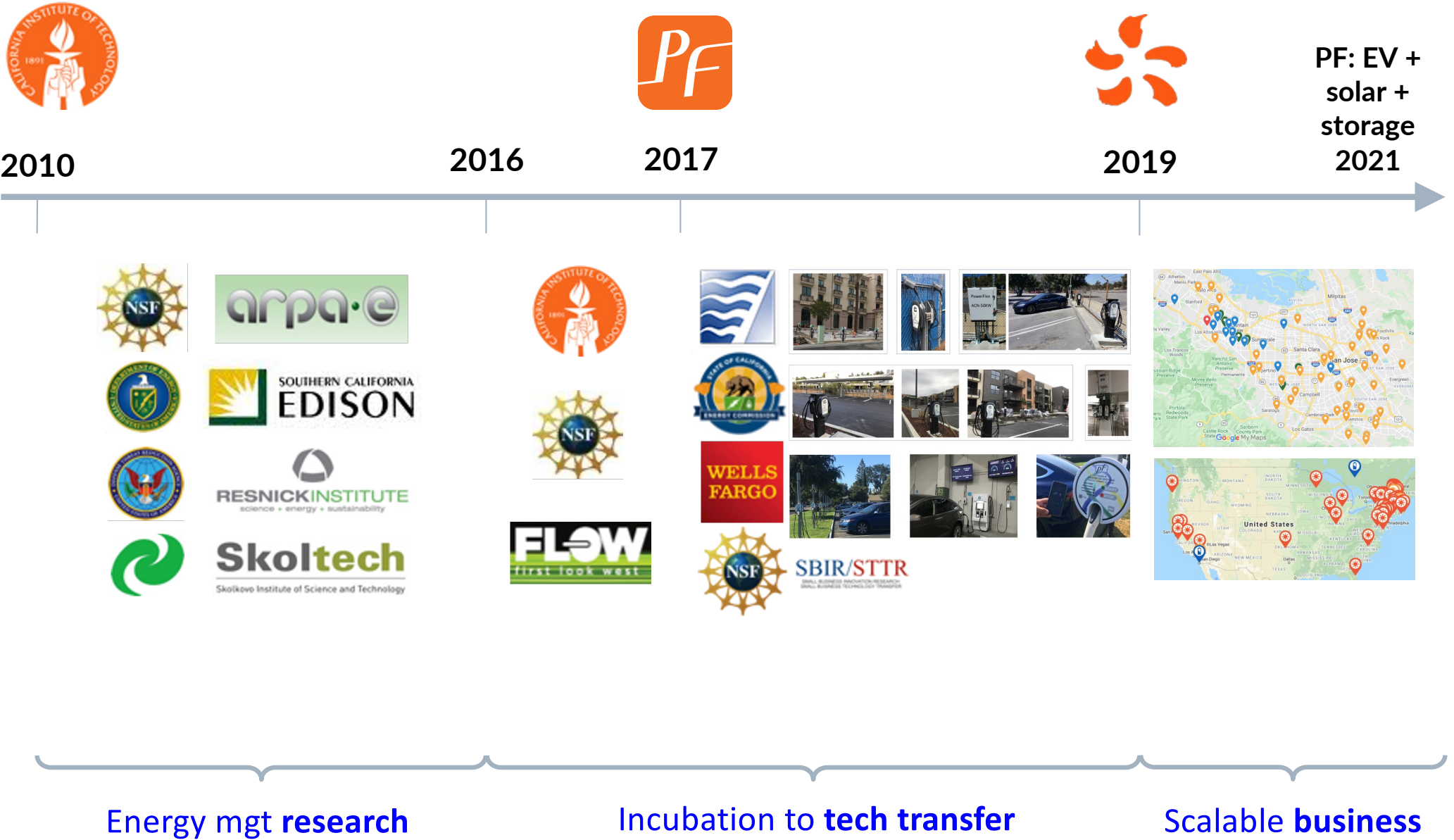


weekdays

weekdays

March 16 Monday, 2020

Commercialization: timeline





Business case: lower capital cost

Table ES.1: Projections for Statewide PEV Charger Demand
Demand for L2 Destination (Workplace and Public) Chargers
(The Default Scenario)

	Total PEVs	Lower Estimate (Chargers)	Higher Estimate (Chargers)
As of 2017	239,328	21,502	28,701
By 2020	645,093	53,173	70,368
By 2025	1,321,371	99,333	133,270

100,000 Chargers @\$15k/ea = \$1.5B

\$15k/charger is unsustainable

CA CEC & IOU incentive program estimated
~\$15k/charger (inc. make ready)

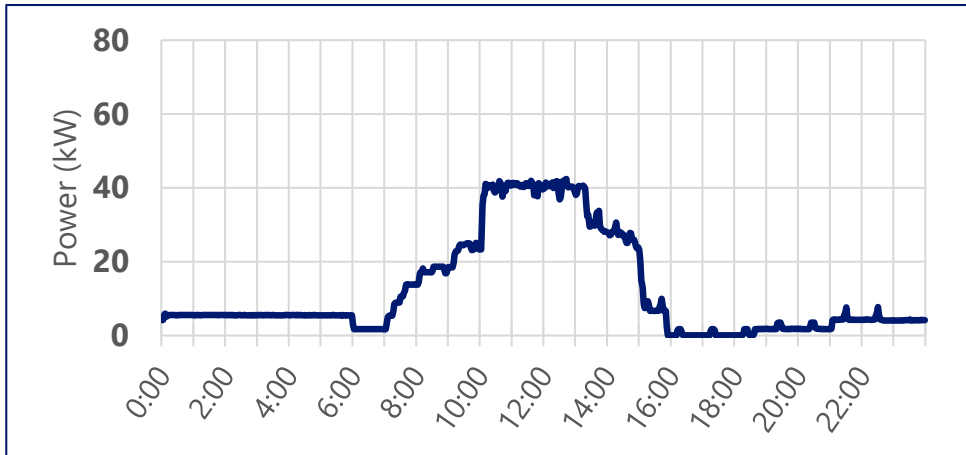
CEC 3/2018 Staff Report

- 168 chargers
 - 118x Universal (J1772) x 6.6kW
 - 50x Tesla x 16kW
- 1.578MW nameplate
 - Connected to 800A/480V panel (max load @80% = 522kW)
 - 3x capacity
 - No Interconnection Upgrade
- Cost: <\$3,000/station

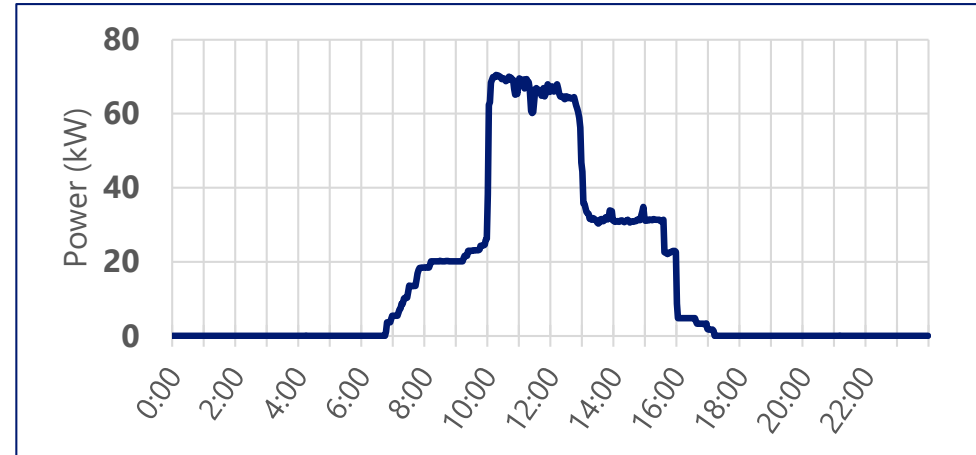
PowerFlex case study: <\$3k/charger
(inc. make ready)



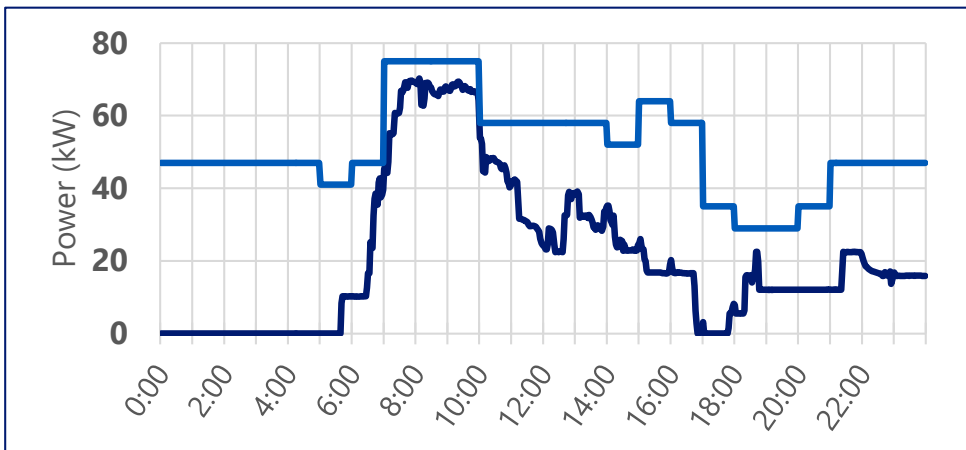
Business case: lower operating cost



Peak Reduction: Reduced Peak by 40% (72kW to 42kW) while still delivering same amount of energy



10am Floodgates: Charging maximized to transformer limits during 10am-2pm to optimize for incentives for consuming surplus solar energy



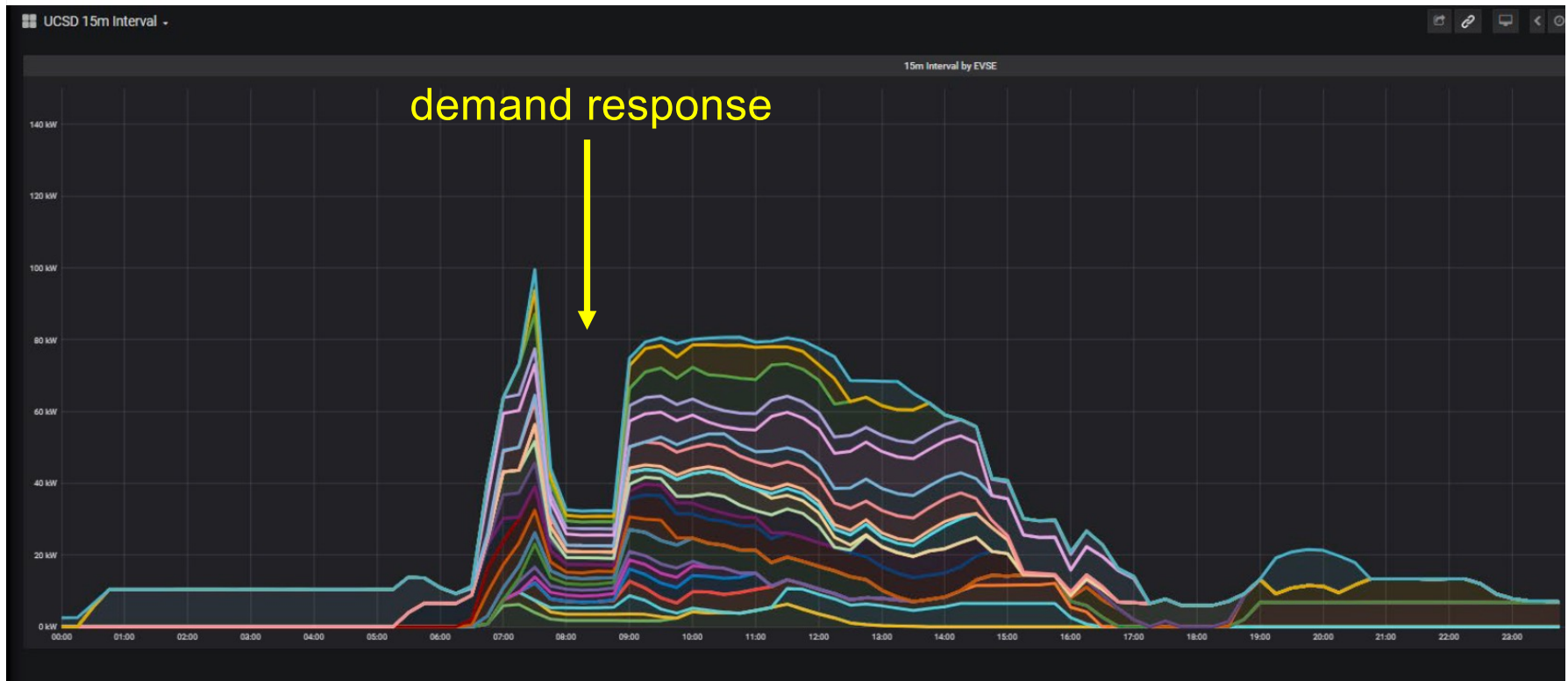
LCFS Curve Following: Charging optimized under LCFS Time-of-Use Value curve

3 ways to reduce operating cost

- Demand charge reduction
- Price arbitrage on ToU tariff
- Increasing LCFS revenue
- EDF – Athena (San Diego, CA)



Business case: grid services





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- Motivation, 3-phase network models





ACN Research Portal

2019 ACM e-Energy:

ACN-Data: Analysis and Applications of an Open EV Charging Dataset

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IEEE TRANSACTIONS ON SMART GRID, VOL. 12, NO. 6, NOVEMBER 2021

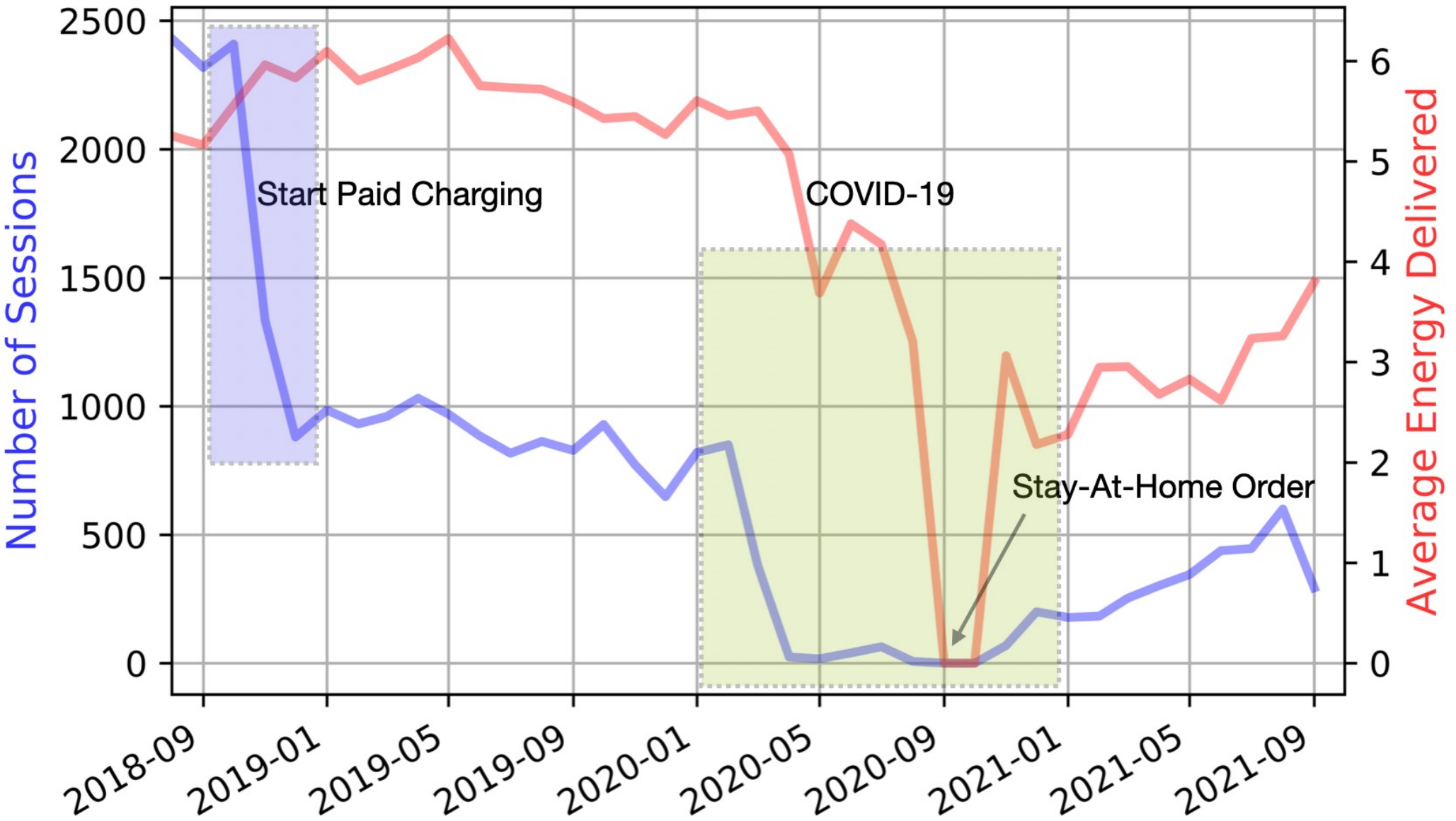
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ACN-Sim: An Open-Source Simulator for Data-Driven Electric Vehicle Charging Research

Zachary J. Lee^{ID}, Sunash Sharma^{ID}, Daniel Johansson, and Steven H. Low^{ID}, *Fellow, IEEE*



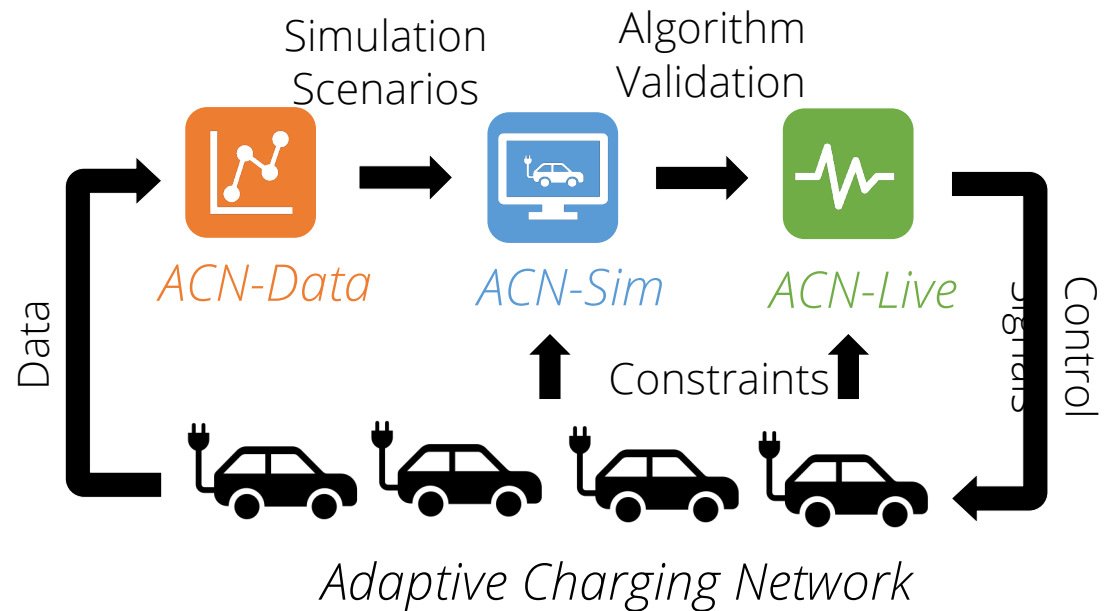
Caltech ACN





ACN research portal

- ACN-Data
- ACN-Sim
- ACN-Live (HW-in-the-loop)



Lee, Li, Low. ACN-Data: analysis and applications of an open EV charging Dataset
ACM e-Energy, June 2019

Lee, Johansson, Low. ACN-Sim: an open-source simulator for data-driven EV charging research
IEEE SmartGridComm, October 2019



ACN-Data

Caltech, JPL, Bay Area office

- 80,000+ EV charging sessions (March 2021)
- Publicly available: ev.caltech.edu
- Growing daily

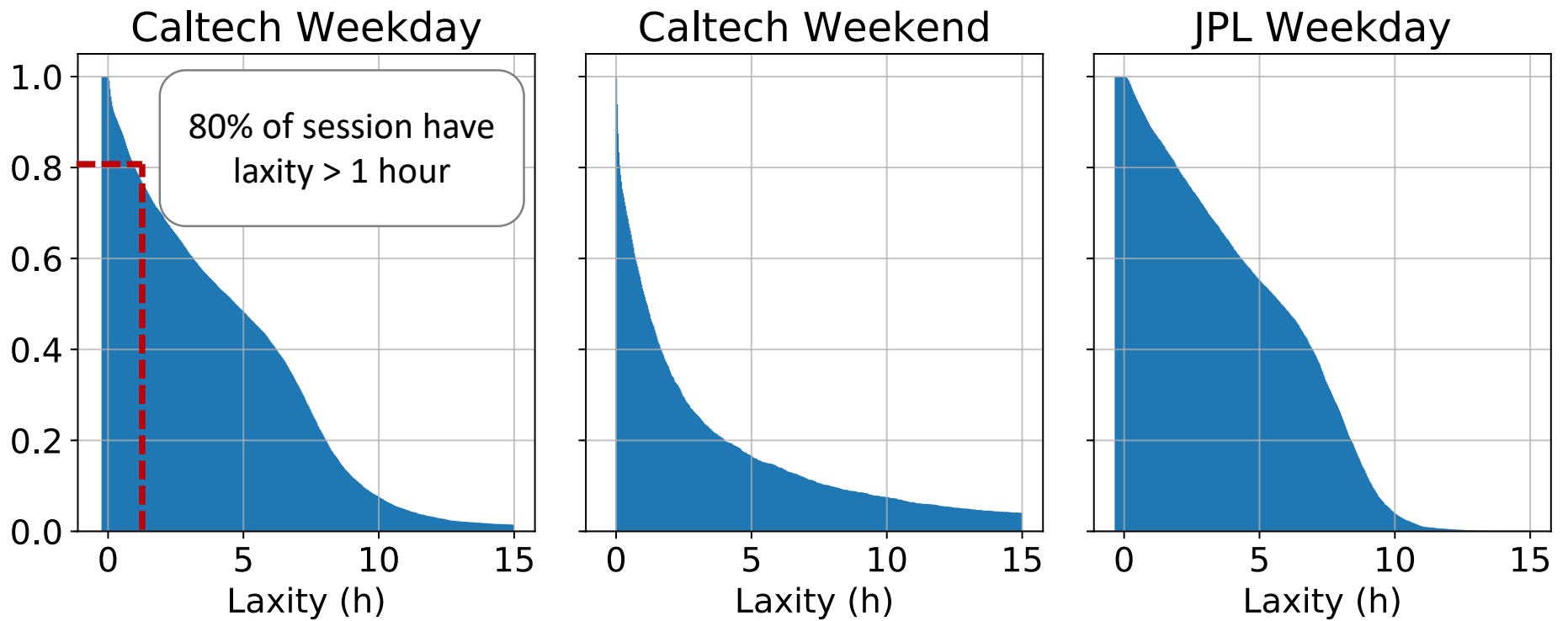
Real **fine-grained** data for

- Modeling user behavior
- Evaluating charging algorithms
- Evaluating charging facilities
- Evaluating grid impacts



User flexibility

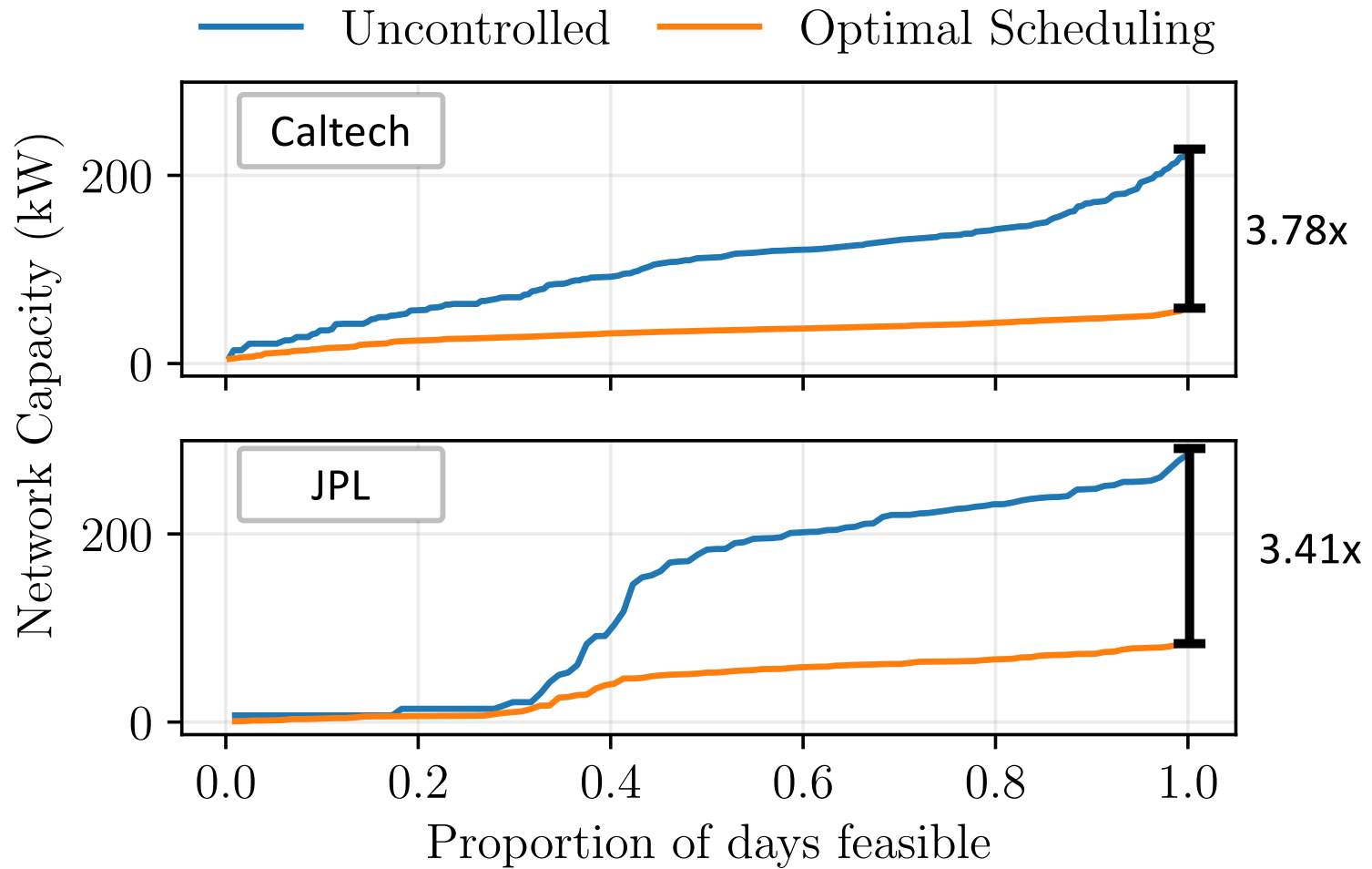
User flexibility



$\text{laxity} := \text{session duration} - \text{min charging time}$



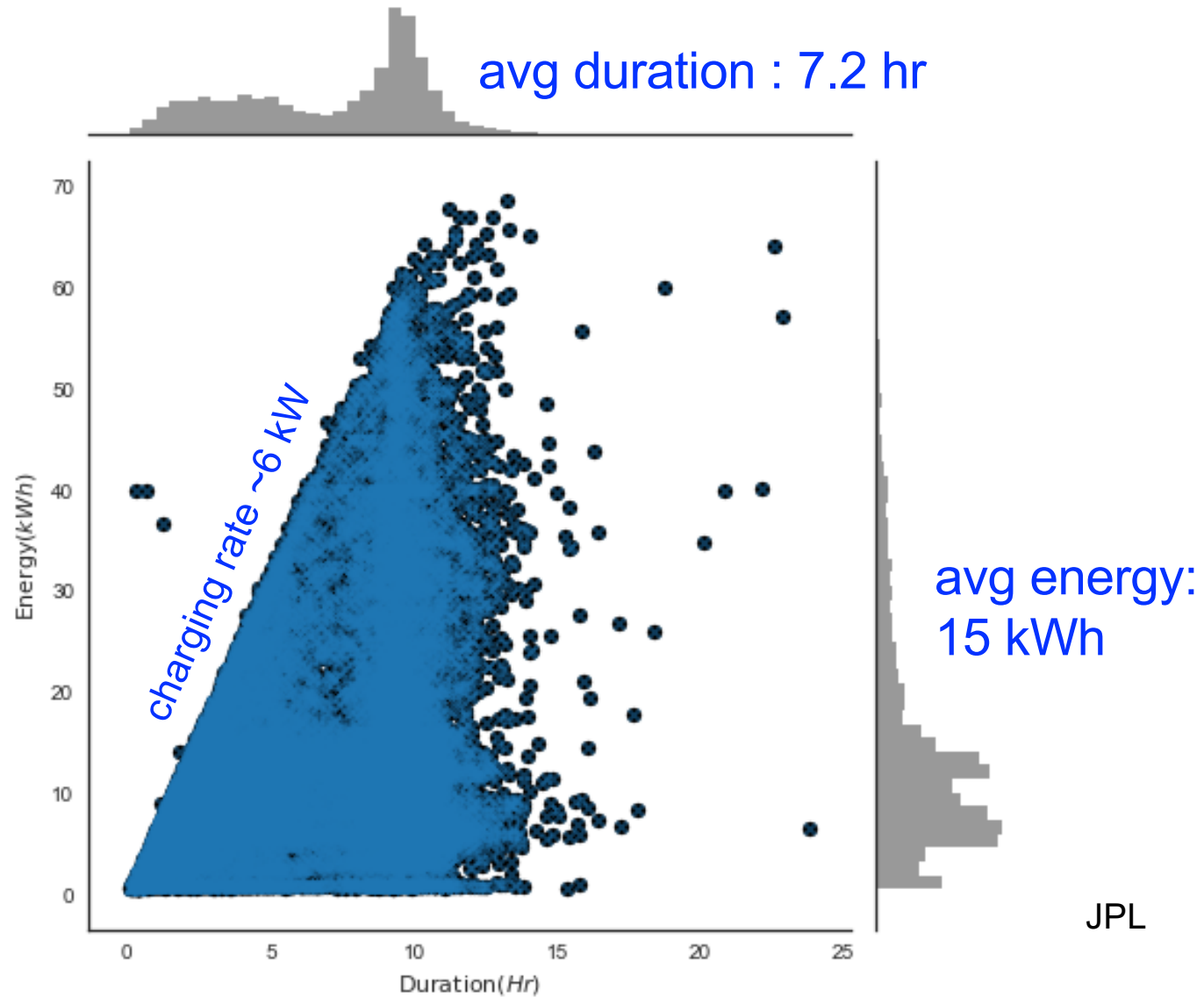
ACN flexibility





User behavior

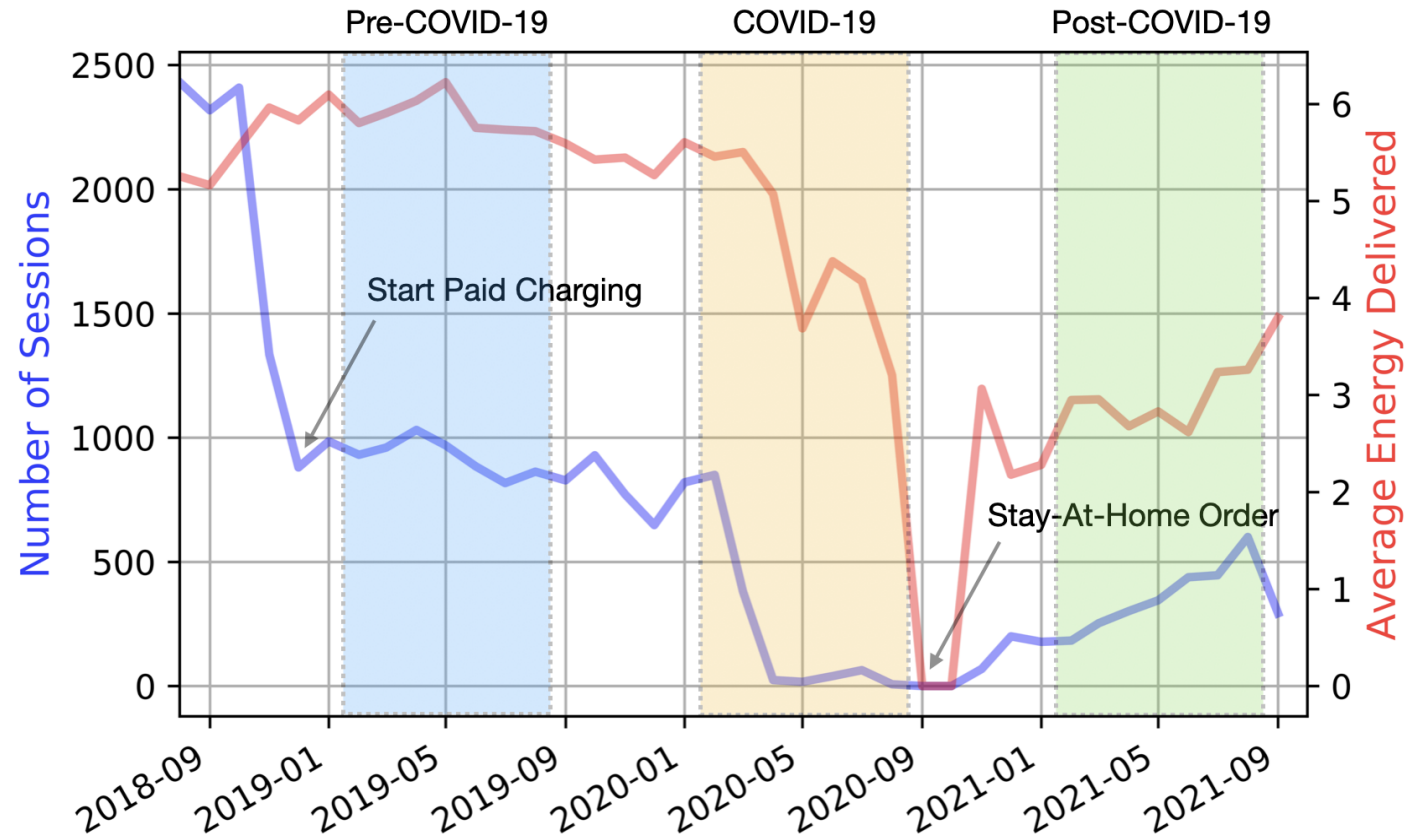
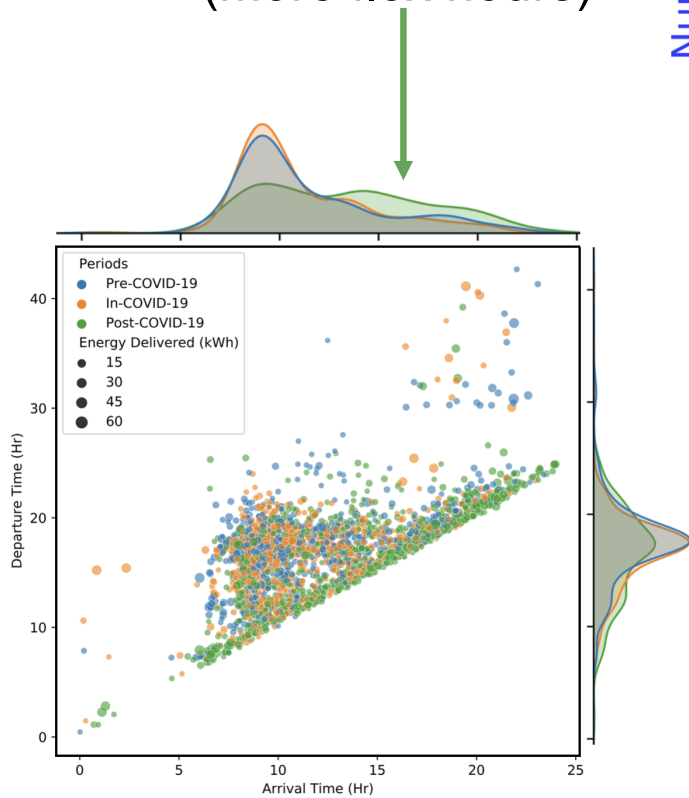
Duration and energy delivered





User behavior

arrival times have larger var post-COVID (more flex hours)



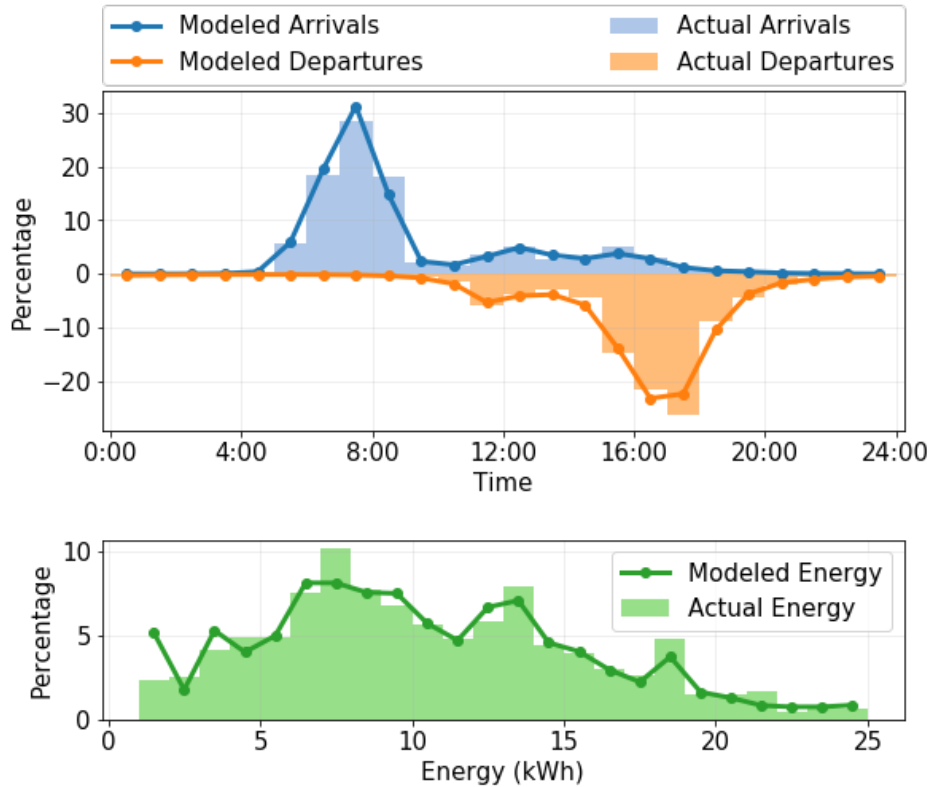
effect of payment & COVID



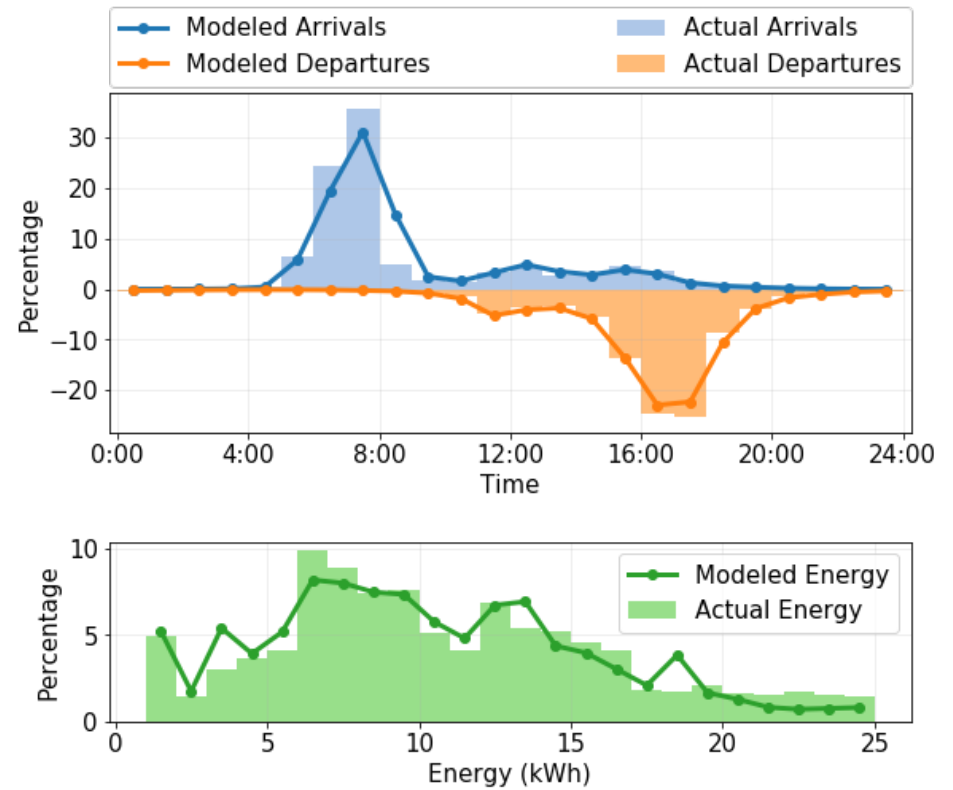
Learning user behavior

Gaussian mixture model

Testing Accuracy (9/1/18 - 11/1/18)

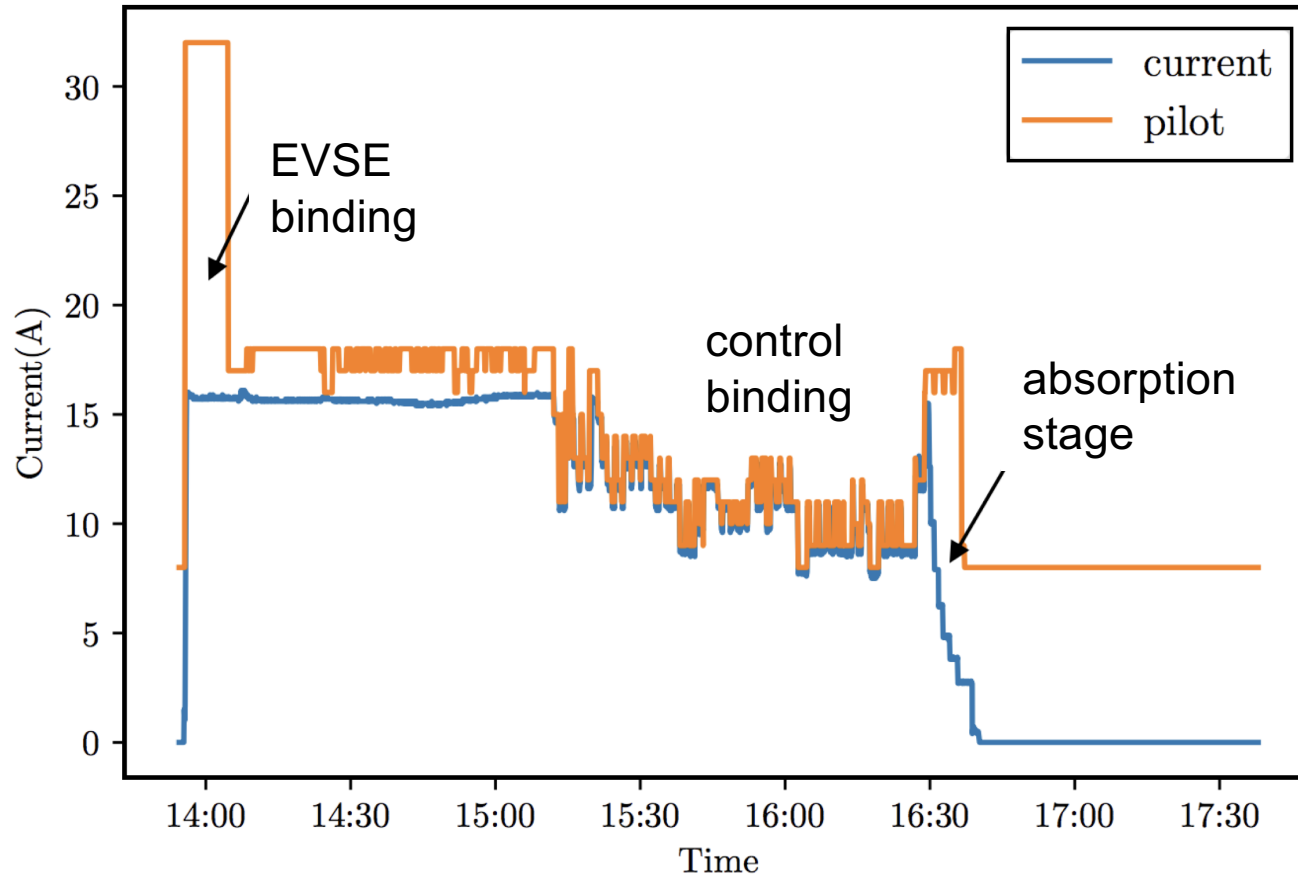


Evaluation Set Accuracy (12/1/18 - 5/1/19)





Charging curves



Caltech Oct 13, 2018

Time series: every 5-10 secs

- pilot signal from controller
- actual current drawn by EV



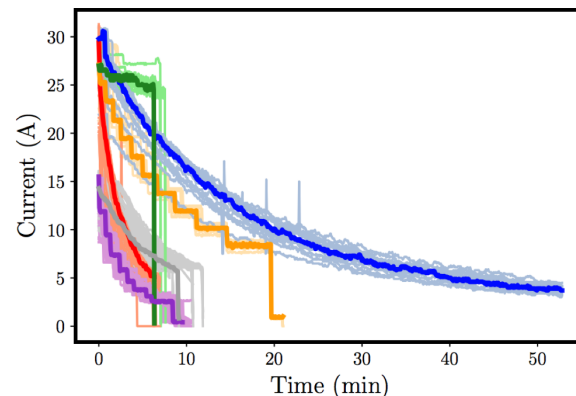
Learning charging curves

Goal: learn representative battery behaviors

- Only small # of batteries used by small # drivers underlying 35,000 charging curves

Challenge: do not know SoC

- Can only characterize tail behavior (absorption stage)
- Charging optimization, BMS actions, missing & noisy data



need to

- extract charging tails
- cluster charging tails

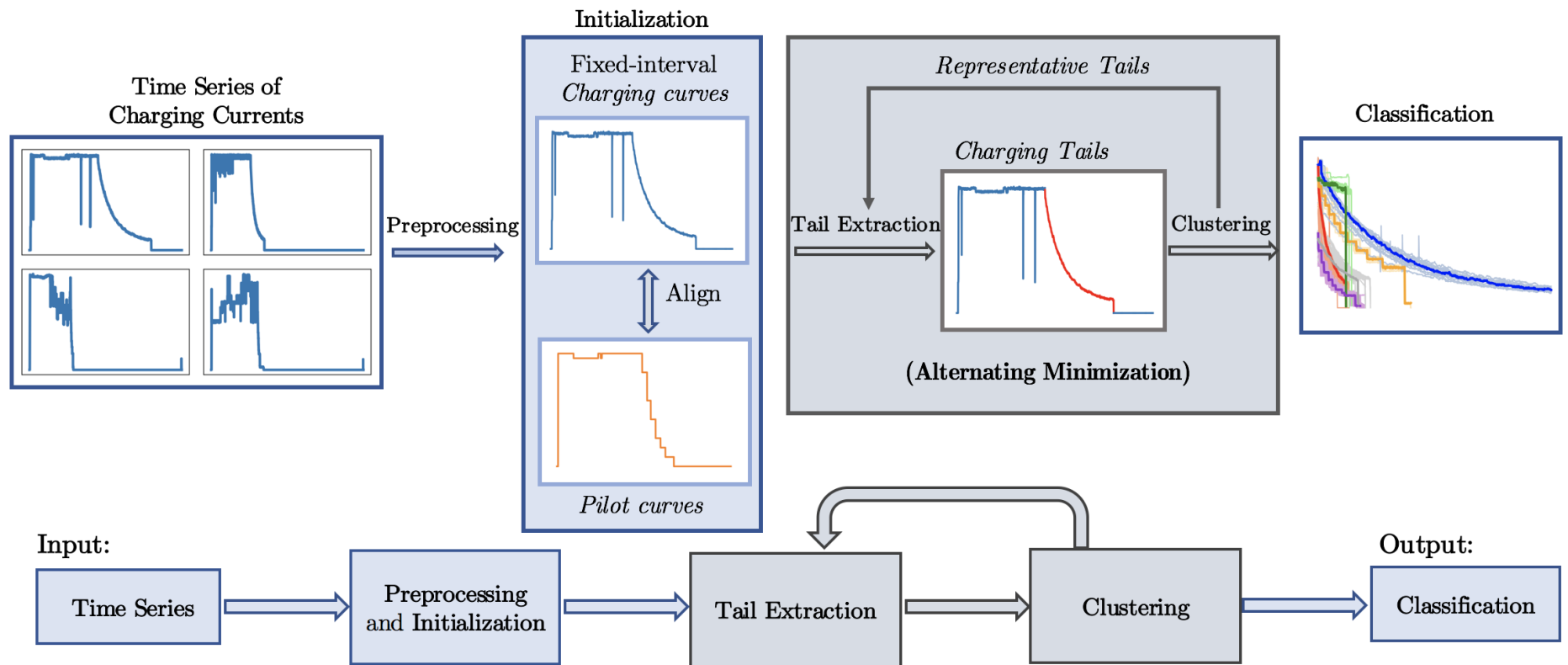
Chenxi Sun, Tongxin Li, S. H. Low and Victor Li.

Classification of EV charging time series with selective clustering

PSCC July 2020



Learning charging curves

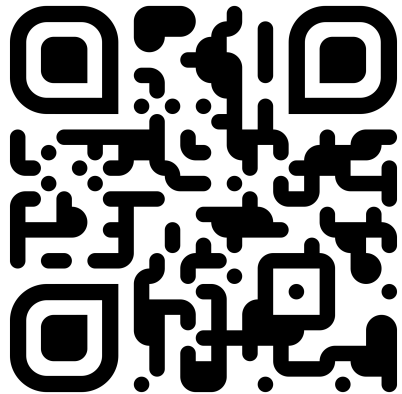


Chenxi Sun, Tongxin Li, S. H. Low and Victor Li.
Classification of EV charging time series with selective clustering
PSCC July 2020




Accessing ACN - Data


- Web Interface
- API
- Python Client
- ACN-Sim



ev.caltech.edu

Site
Caltech

From
01/01/2019 12:00 AM 

To
06/20/2019 9:58 AM 

Minimum Energy (kWh)
5

Sessions Found:
3039

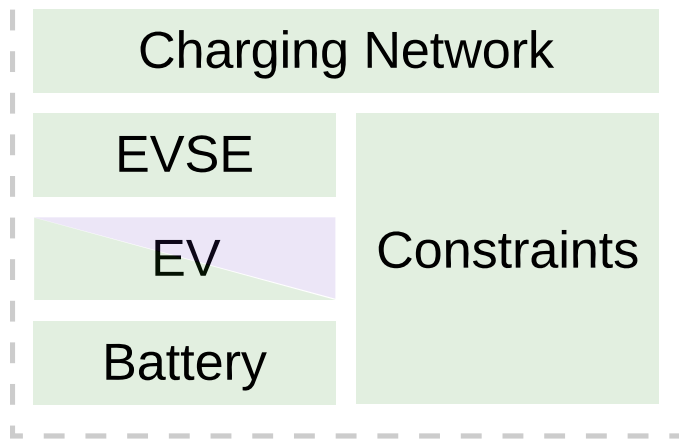
[DOWNLOAD](#)

Caltech

open-source & extensible



ACN - Sim

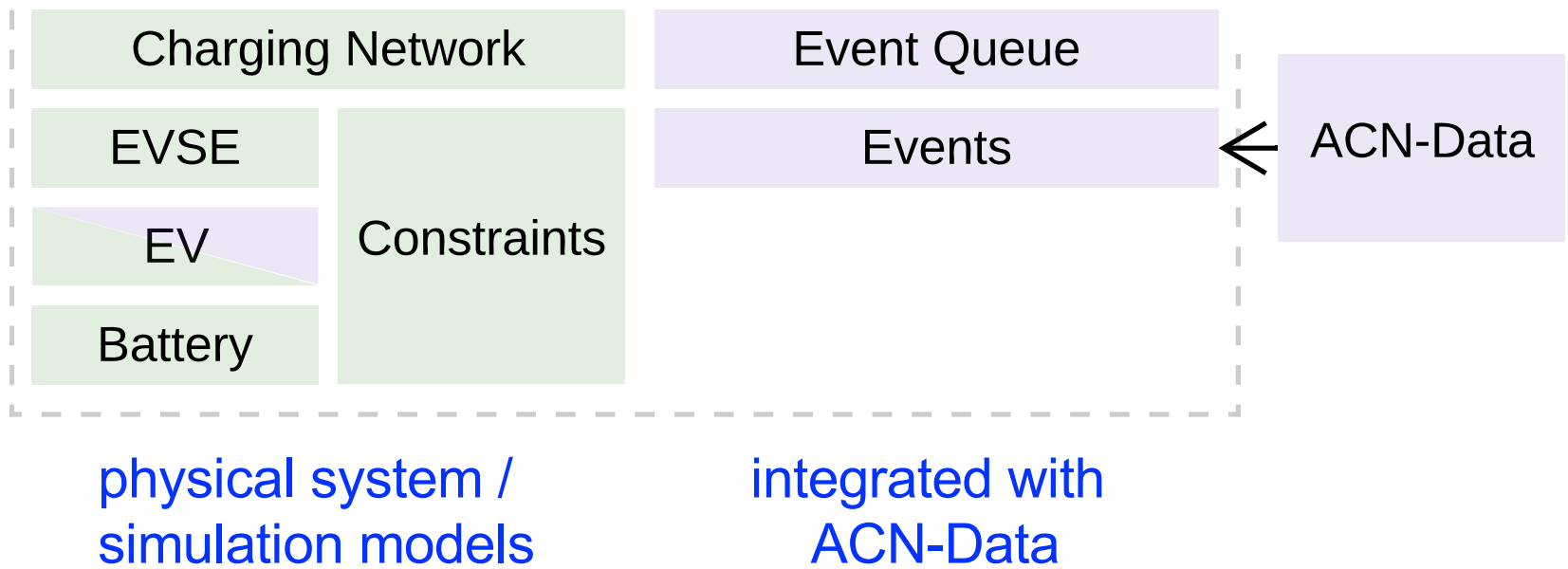


physical system /
simulation models

open-source & extensible



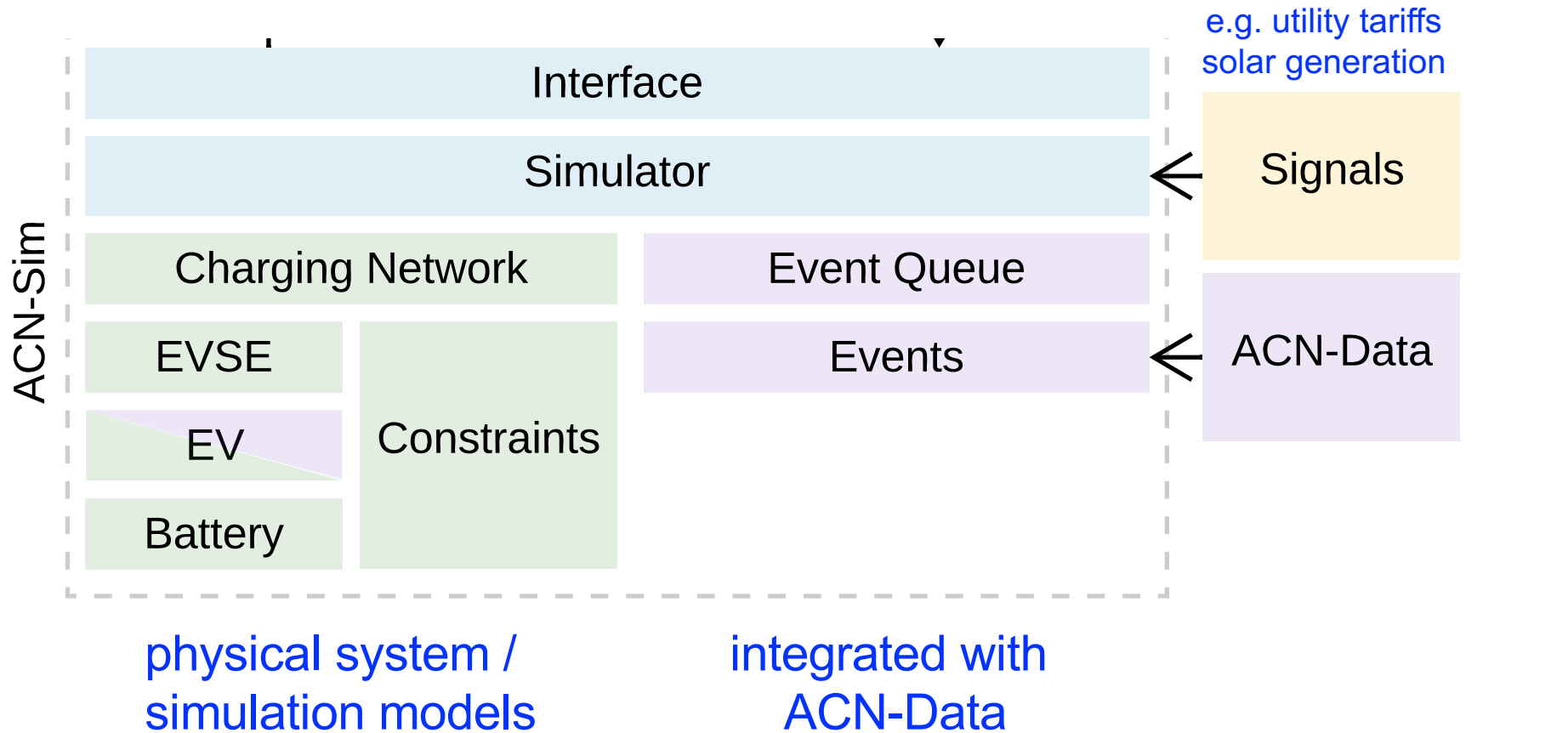
ACN - Sim



open-source & extensible



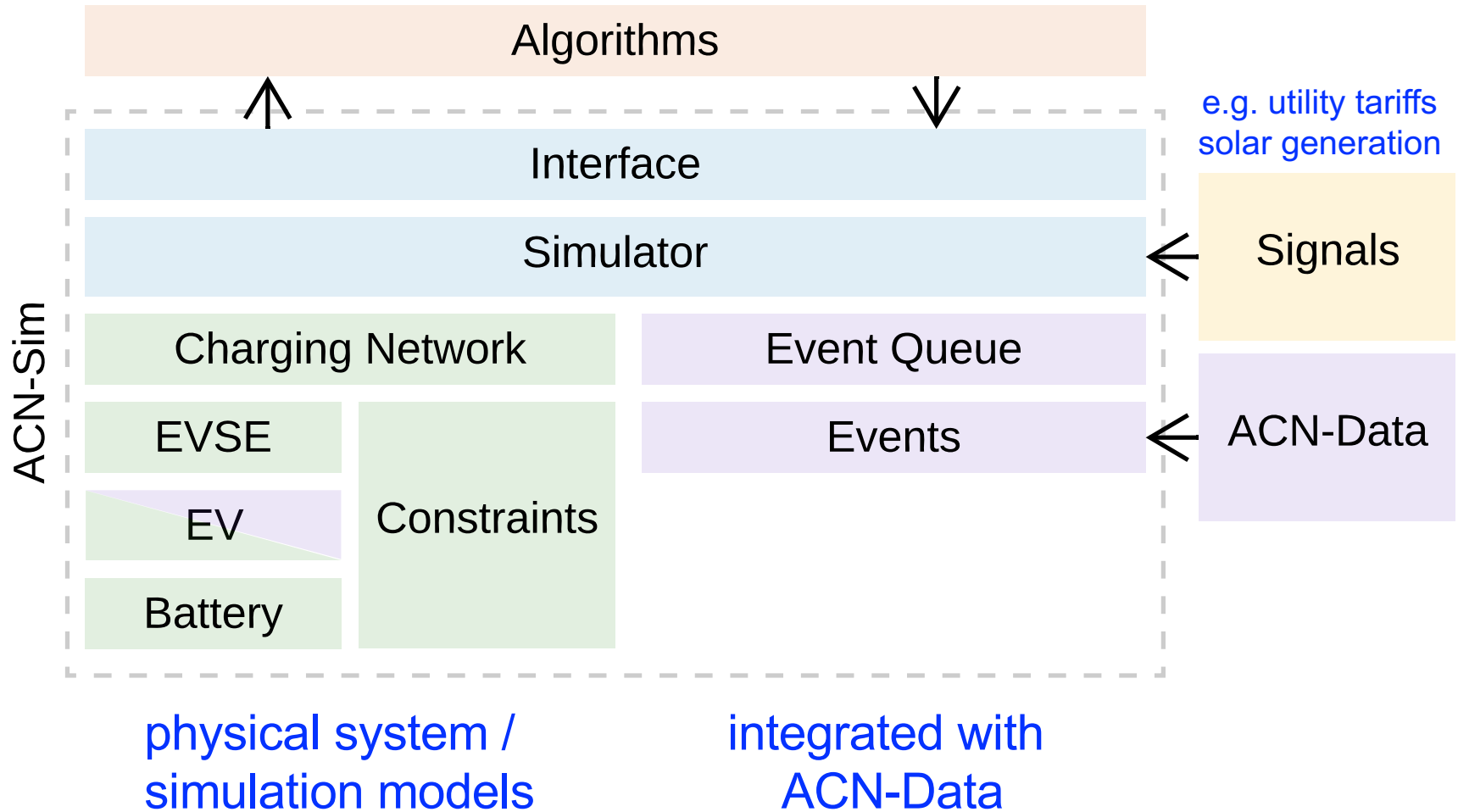
ACN - Sim



open-source & extensible



ACN - Sim



open-source & extensible



Grid impact

How can large-scale EV charging mitigate Duck Curve ?



Charging model

N EVs: $i = 1, \dots, N$
 T control intervals: $t = 1, \dots, T$
 EV i : $(e_i, a_i, d_i, \bar{r}_i)$

energy demand (miles / kWh)
 arrival / departure time
 peak charging rate (kW)

customizable utility functions

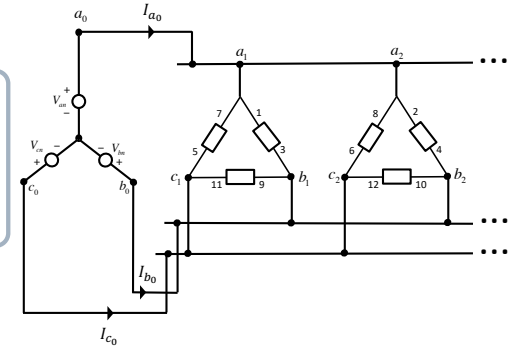
$$\max_r \sum_v \alpha_v u_v(r)$$

Compute: charging rates

$$r := (r_i(t), i = 1, \dots, N, t = 1, \dots, T)$$

$$0 \leq r_i(t) \leq \bar{r}_i(t)$$

$$\sum_{t \in T} r_i(t) \leq e_i$$



Ckt No	BUS			Ckt No
	A	B	C	
1				2
3				4
5				6
7				8
9				10
11				12
13				14
15				16
17				18
19				20
21				22
23				24
25				26
27				28
29				30
31				32
33				34
35				36
37				38
39				40
41				42

infrastructure constraints

$$\left| \sum_{i \in \mathcal{V}} A_{li} r_i(t) e^{j\phi_i} \right| \leq c_{lt}(t)$$

SoC constraints, or linear approx.

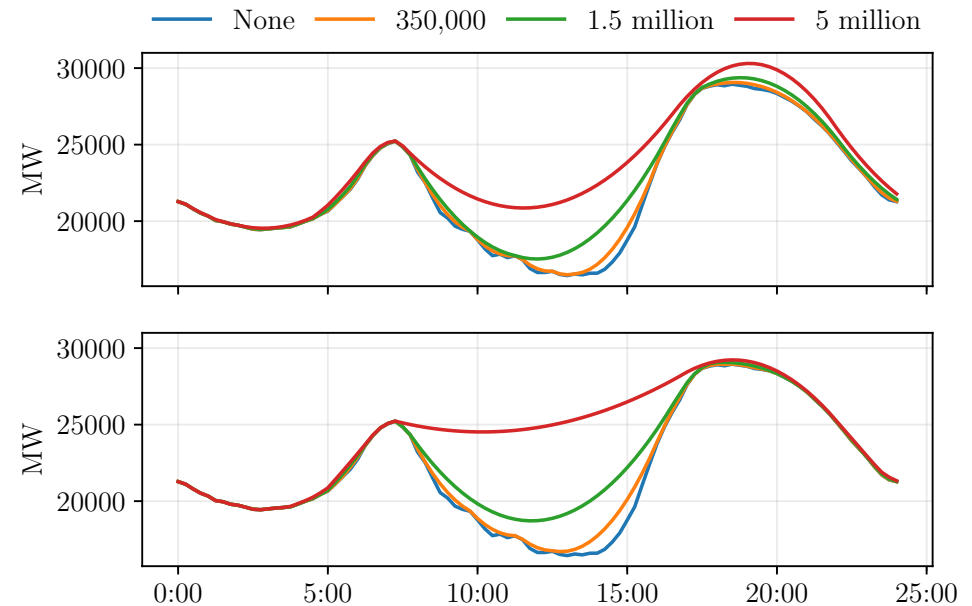


Grid impact

MPC

$$\begin{aligned} \max_r \quad & \sum_v \alpha_v u_v(r) \\ \text{subject to} \quad & 0 \leq r_i(t) \leq \bar{r}_i(t) \\ & \sum_{t \in \mathcal{T}} r_i(t) \leq e_i \\ & \left| \sum_{i \in \mathcal{V}} A_{li} r_i(t) e^{j\phi_i} \right| \leq c_{lt}(t) \end{aligned}$$

MPC in real system is a lot more

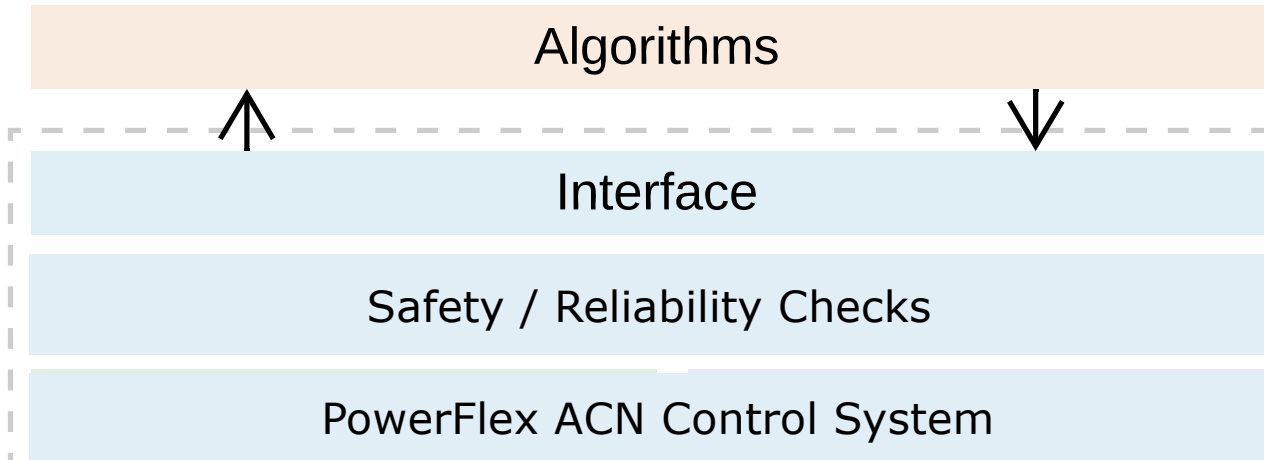


Minimize evening ramp based on **real data**

- EV data from ACN-Data
- Simulation models from ACN-Sim
- CAISO solar and load data
- Simple estimate without grid model



ACN - Live



open-source
& extensible



ACN research portal

Adaptive Charging Network

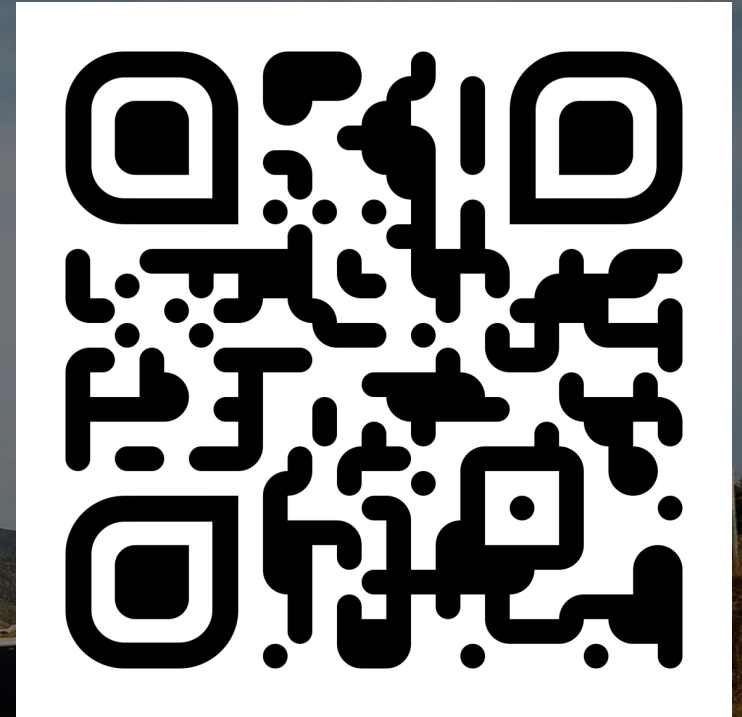
HOME INFO RESEARCH DATA SIMULATOR ACCOUNT ▾

The Adaptive Charging Network

Accelerating Electric Vehicle Research @ Caltech and Beyond

zlee@caltech.edu

ev.caltech.edu





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ACN Pricing



ELSEVIER

Electric Power Systems Research

Volume 189, December 2020, 106694



Pricing EV charging service with demand charge ☆

Zachary J. Lee ^a  , John Z.F. Pang ^b , Steven H. Low ^{a, b} 

PSCC 2020



Online adaptive charging

Model predictive control:

$$\begin{aligned} \max_r \quad & \sum_v \alpha_v u_v(r) \\ \text{subject to} \quad & 0 \leq r_i(t) \leq \bar{r}_i(t) \\ & \sum_{t \in \mathcal{T}} r_i(t) \leq e_i \\ & \left| \sum_{i \in \mathcal{V}} A_{li} r_i(t) e^{j\phi_i} \right| \leq c_{lt}(t) \end{aligned}$$



Pricing design

Charging design

- Must adapt to system state in real time
- Objectives must be customized for site hosts

Pricing design: recover cost for site hosts

- Energy
- Externality: system peak (demand charge)
- Externality: infrastructure congestion

Key idea: **decouple** charging and pricing

- Drivers receive energy in time, at **minimum** payments
- Charging is **socially** optimized by MPC
- Site host fully **recovers** electricity cost

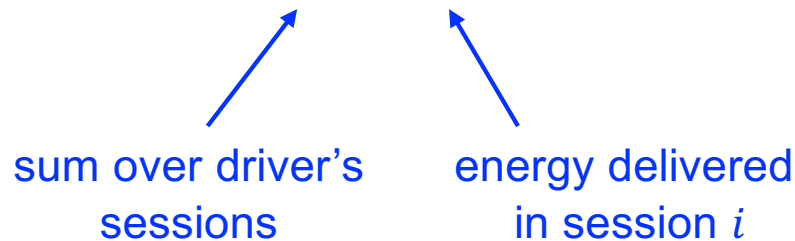


Offline optimal pricing

start with conclusion ...

At end of month

- Compute **ex post session** price α_i^*
- Driver pays: $\sum_i \alpha_i^* e_i$



No uncertainty nor need for ToU tariff or demand forecasts



Pricing design

$$C(r) := \sum_t p_t \sum_i r_i(t) + P \max_t \underbrace{\sum_i r_i(t)}_{\text{peak power}}$$

time-varying tariff \$/kWh

demand charge \$/kWh

1. What is min **system electricity** cost to meet demand ?
2. How to **fairly** allocate system cost to drivers ?



Pricing design

$$C(r) := \sum_t p_t \sum_i r_i(t) + P \max_t \sum_i r_i(t)$$

Pricing min **system** cost:

$$\begin{aligned} C^{\min} &:= \min && \sum_t p_t \sum_i r_i(t) + Pq \\ &&& \text{s. t.} && \sum_t r_i(t) = e_i, && \text{meet demand} && \alpha_i \\ &&& && \sum_i A_{li} r_i(t) \leq c_{lt} && \text{infrastructure} && \beta_{lt} \\ &&& && && \text{capacity limit} && \\ &&& && r_i(t) \leq \bar{r}_i(t), && \text{EVSE limit} && \gamma_{it} \\ &&& && q \geq \sum_i r_i(t), && \text{system peak} && \delta_t \end{aligned}$$



Pricing design

Fairly (incentive compatibly) allocate system cost to EVs

$$\pi_i^*(t) := \underbrace{p_t}_{\text{energy}} +$$

time-varying
tariff



Pricing design

Fairly (incentive compatibly) allocate system cost to EVs

$$\pi_i^*(t) := \underbrace{p_t}_{\text{energy}} + \underbrace{\sum_l A_{li} \beta_{lt}^*}_{\text{network congestion}} + \underbrace{\gamma_{it}^*}_{\text{charger congestion}} + \underbrace{\delta_t^*}_{\text{demand charge}}$$

- Driver & time dependent prices

Driver pays for each session i

$$\Pi_i^* := \sum_t \pi_i^*(t) r_i^*(t)$$

This achieves pricing goals: recovers

- Energy cost
- Congestion rents
- Demand charge EV i is responsible for



Pricing design

Design principle:

$$\pi_i^*(t) := \underbrace{p_t}_{\text{energy}} + \underbrace{\sum_l A_{li} \beta_{lt}^*}_{\text{network congestion}} + \underbrace{\gamma_{it}^*}_{\text{charger congestion}} + \underbrace{\delta_t^*}_{\text{demand charge}}$$

$$\Pi_i^* = \sum_t \pi_i^*(t) r_i^*(t)$$

Theorem

1. Demand charge: $P = \sum_t \delta_t^*$ EVs that cause peak will pay



Pricing design

Design principle:

$$\pi_i^*(t) := \underbrace{p_t}_{\text{energy}} + \underbrace{\sum_l A_{li} \beta_{lt}^*}_{\text{network congestion}} + \underbrace{\gamma_{it}^*}_{\text{charger congestion}} + \underbrace{\delta_t^*}_{\text{demand charge}}$$

$$\Pi_i^* = \sum_t \pi_i^*(t) r_i^*(t)$$

Theorem

1. Demand charge: $P = \sum_t \delta_t^*$ EVs that cause peak will pay

2. Time-invariant session price α_i^* : $\Pi_i^* = \alpha_i^* e_i$

$\pi_i^*(t) \geq \alpha_i^*$ with $\pi_i^*(t) = \alpha_i^*$ if $r_i^*(t) > 0$ EVs pay min cost



Pricing design

Design principle:

$$\pi_i^*(t) := \underbrace{p_t}_{\text{energy}} + \underbrace{\sum_l A_{li} \beta_{lt}^*}_{\text{network congestion}} + \underbrace{\gamma_{it}^*}_{\text{charger congestion}} + \underbrace{\delta_t^*}_{\text{demand charge}}$$

$$\Pi_i^* = \sum_t \pi_i^*(t) r_i^*(t)$$

Theorem

1. Demand charge: $P = \sum_t \delta_t^*$ EVs that cause peak will pay

2. Time-invariant session price α_i^* : $\Pi_i^* = \alpha_i^* e_i$

$\pi_i^*(t) \geq \alpha_i^*$ with $\pi_i^*(t) = \alpha_i^*$ if $r_i^*(t) > 0$ EVs pay min cost

3. Cost recovery: $\sum_i \Pi_i^* \geq C^{min}$

$$\sum_i \Pi_i^* - C^{min} = \sum_{t,l} c_{lt} \beta_{lt}^* + \sum_{t,i} \bar{r}_i(t) \gamma_{it}^* \quad \text{Congestion rents}$$



Offline optimal pricing

At end of month

- Compute **ex post session** price α_i^*
- Driver pays: $\sum_i \alpha_i^* e_i$

No uncertainty nor need for ToU tariff or demand forecasts



Agenda

ACN: Caltech testbed

- Testbed to commercial deployment

ACN: Research Portal

- Data, Sim, Live

ACN: pricing demand charge

- Monthly billing at workplaces

Unbalanced 3-phase modeling

- Motivation, 3-phase network models





Unbalance 3-phase modeling

Power System Analysis

A Mathematical Approach

Steven H. Low

DRAFT available at: <http://netlab.caltech.edu/book/>

Corrections, questions, comments appreciated!



Motivation

Most papers implicitly assume single-phase

- Balanced 3-phase systems have single-phase equivalents

Single-phase models applicable for most purposes

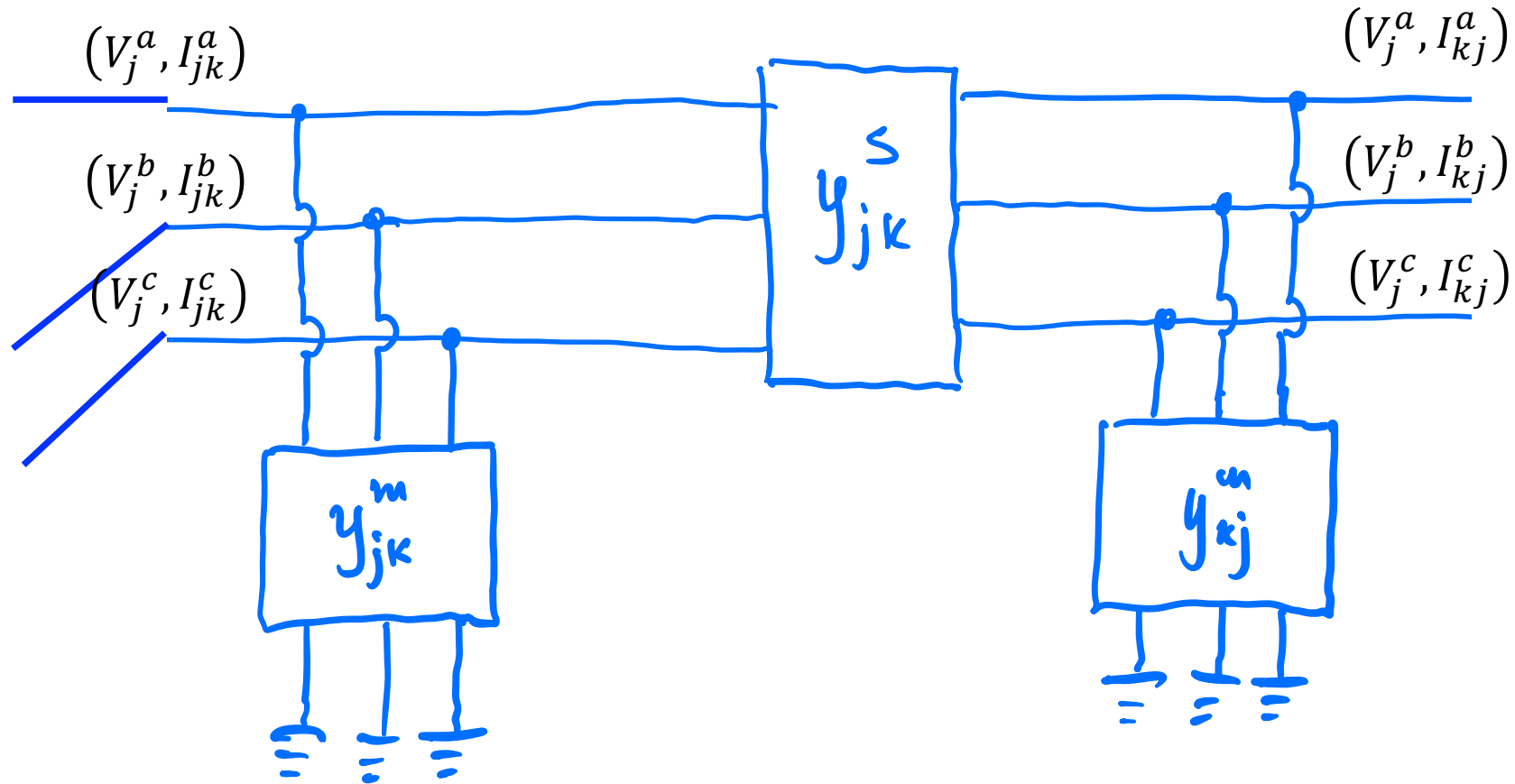
- Transmission system applications
- For illustrating **basic ideas** and analysis of most algorithms (unbalanced 3-phase models structurally similar to 1-phase models)

Unbalanced 3-phase modeling needed

- When control & optimization are explicitly on single-phase devices making up a 3-phase devices
- For implementation in real systems when phases are not balanced



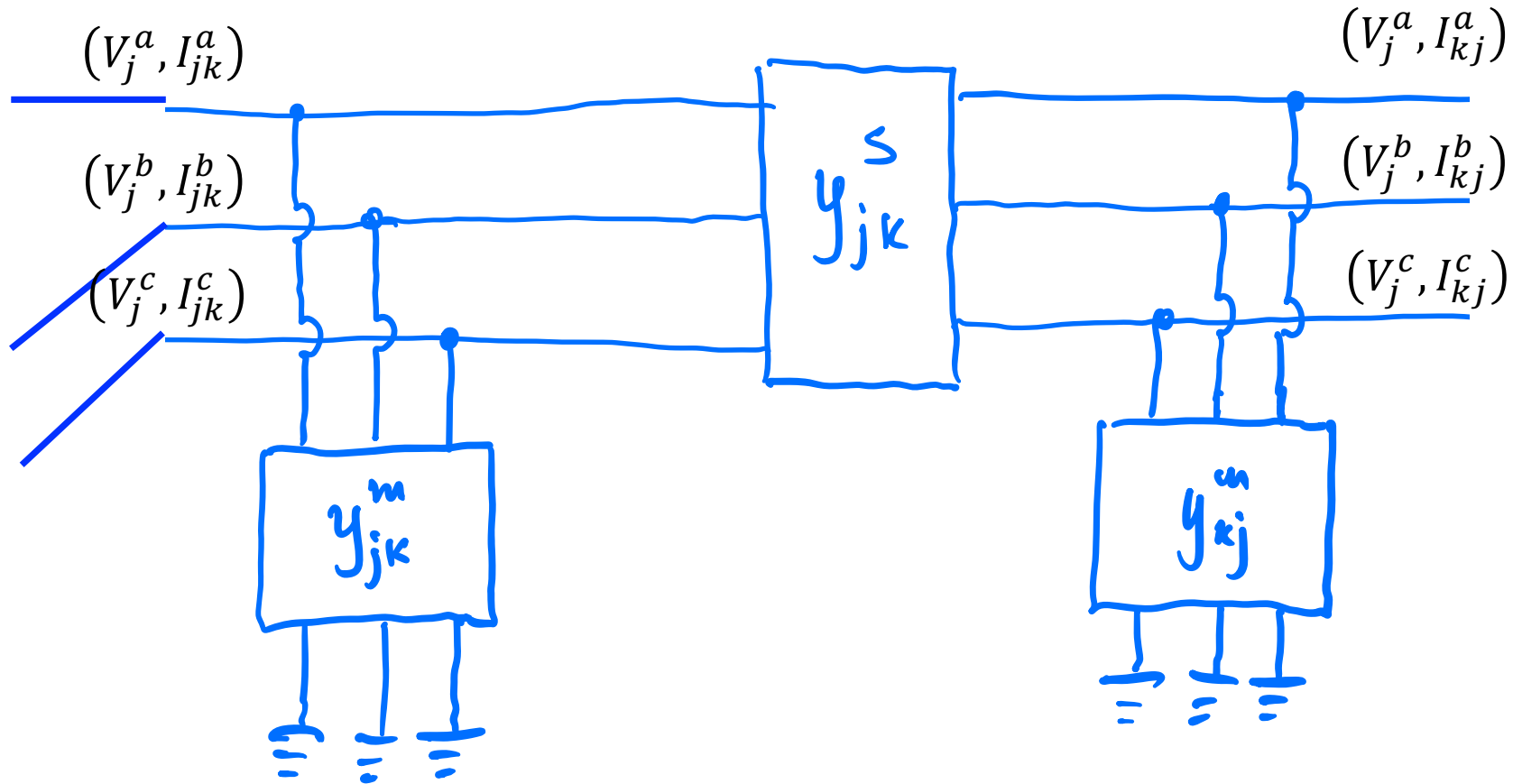
Motivation



- Many models assume **terminal** currents $(I_{jk}^a, I_{jk}^b, I_{jk}^c)$ are controllable (optimization vars)
- Extension to 3-phase setting is straightforward:



Motivation



$$I_{jk} = y_{jk}^s (V_j - V_k) + y_{jk}^m V_j$$

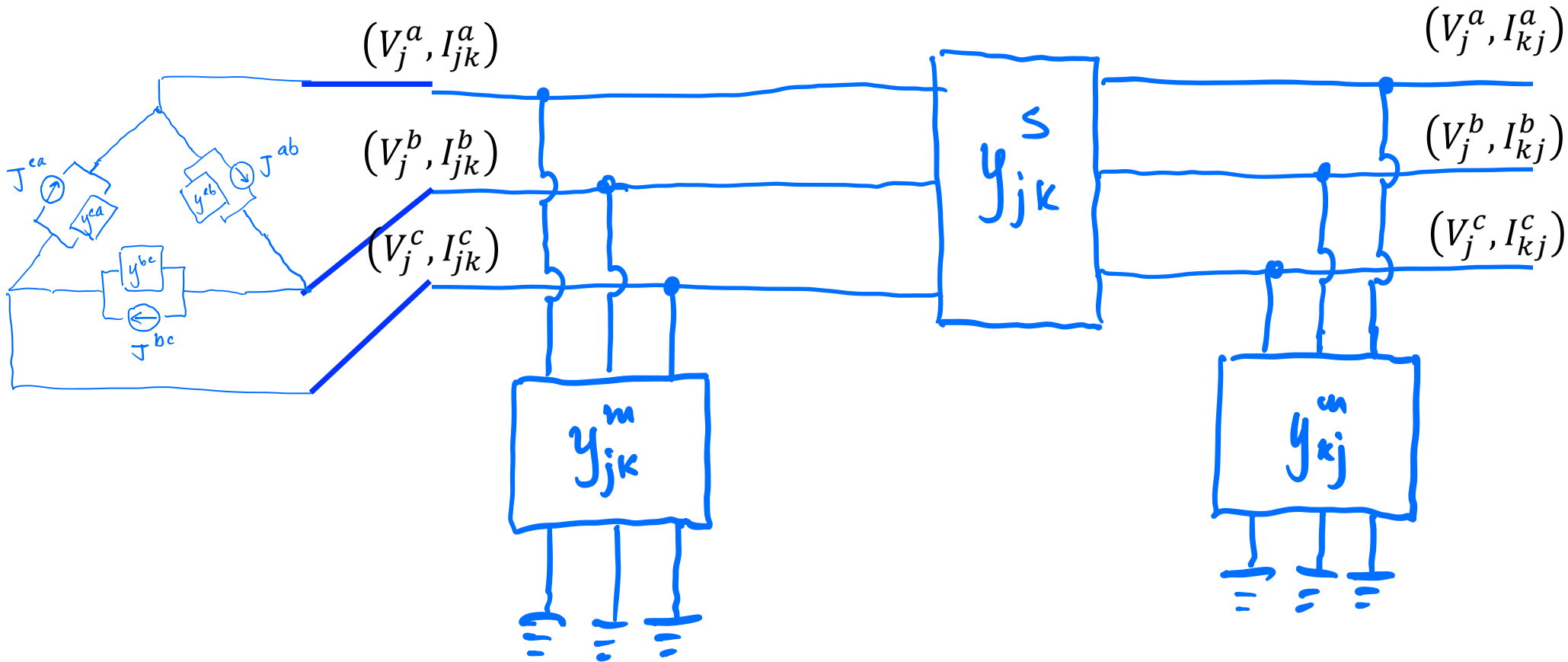
$$I_{kj} = y_{jk}^s (V_k - V_j) + y_{kj}^m V_k$$

1-phase: $I_{jk}, V_j^a \in \mathbb{C} \cdot y_{jk}^{s/m} \in \mathbb{C}$

3-phase: $I_{jk}, V_j^a \in \mathbb{C}^3 \cdot y_{jk}^{s/m} \in \mathbb{C}^{3 \times 3}$



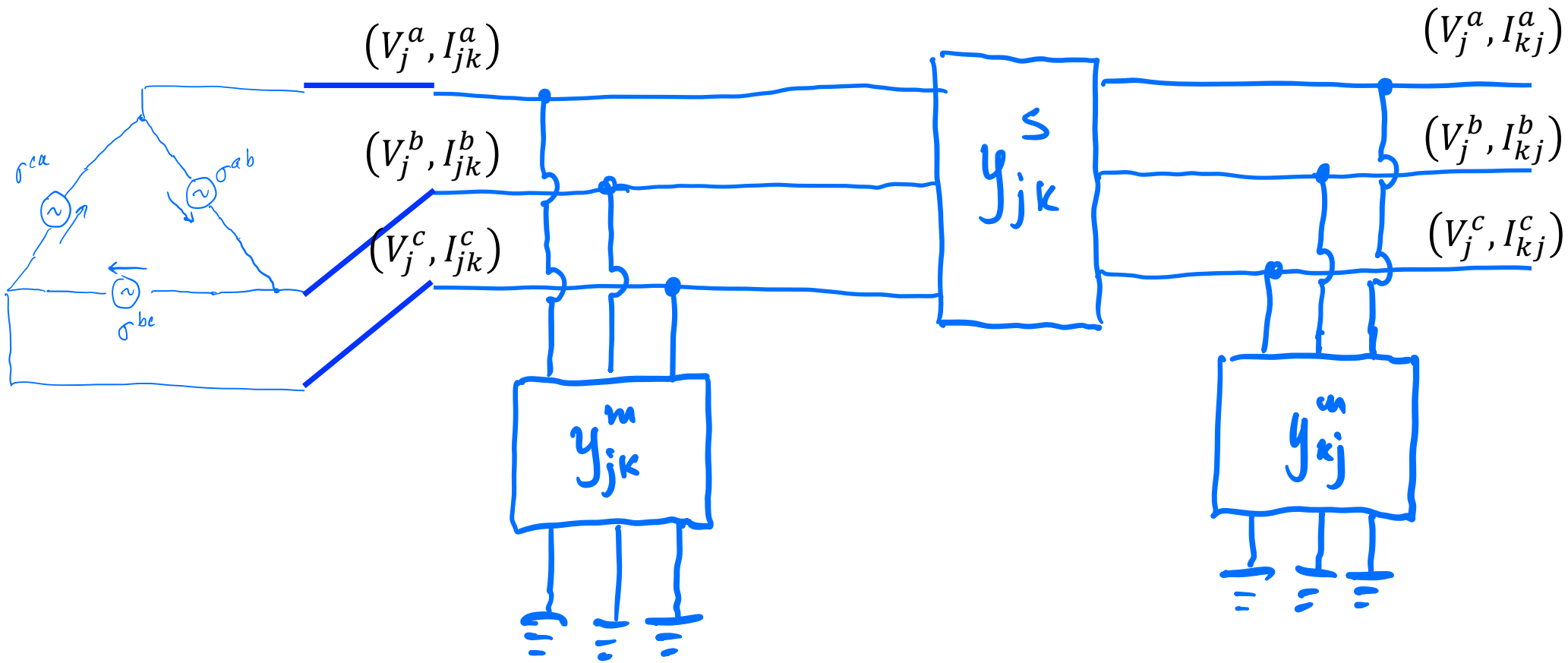
Motivation



- **Terminal** currents I_{jk} are externally observable, but often not directly controllable
- If only **internal** currents $(J_j^{ab}, J_j^{bc}, J_j^{ca})$ of current source are directly controllable, then need a 3-phase device model to convert between internal & terminal vars



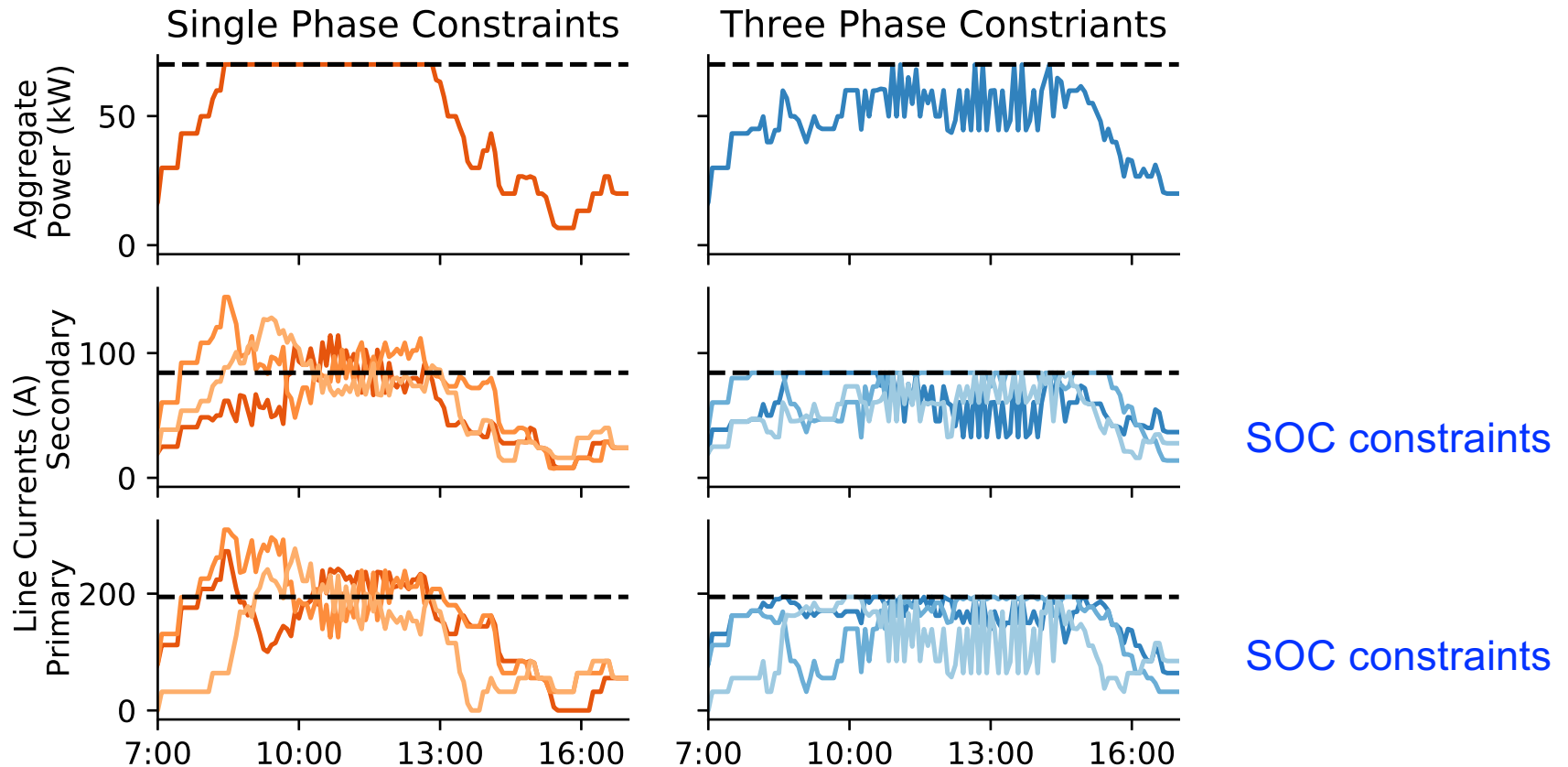
Motivation



Similarly for power sources or voltage sources



Motivation: example

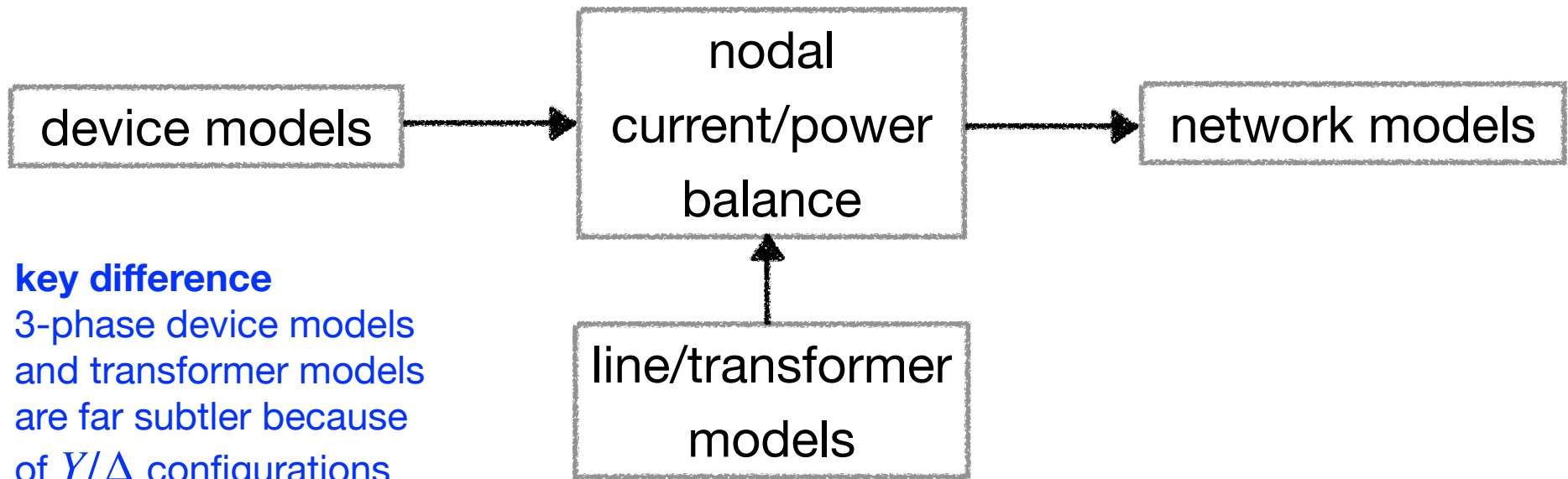


Left panel: Actual 3-phase currents violate capacity constraints if “single-phase constraints” are used (ACN-Sim based on Caltech ACN on Sept 5, 2018 data)

“single-phase constraints” : $\sum_i r_i(t) \leq R$ (no phase line constraints for lack of phase info)



Overview: 3-phase modeling



key difference
3-phase device models
and transformer models
are far subtler because
of Y/Δ configurations

single-phase or 3-phase



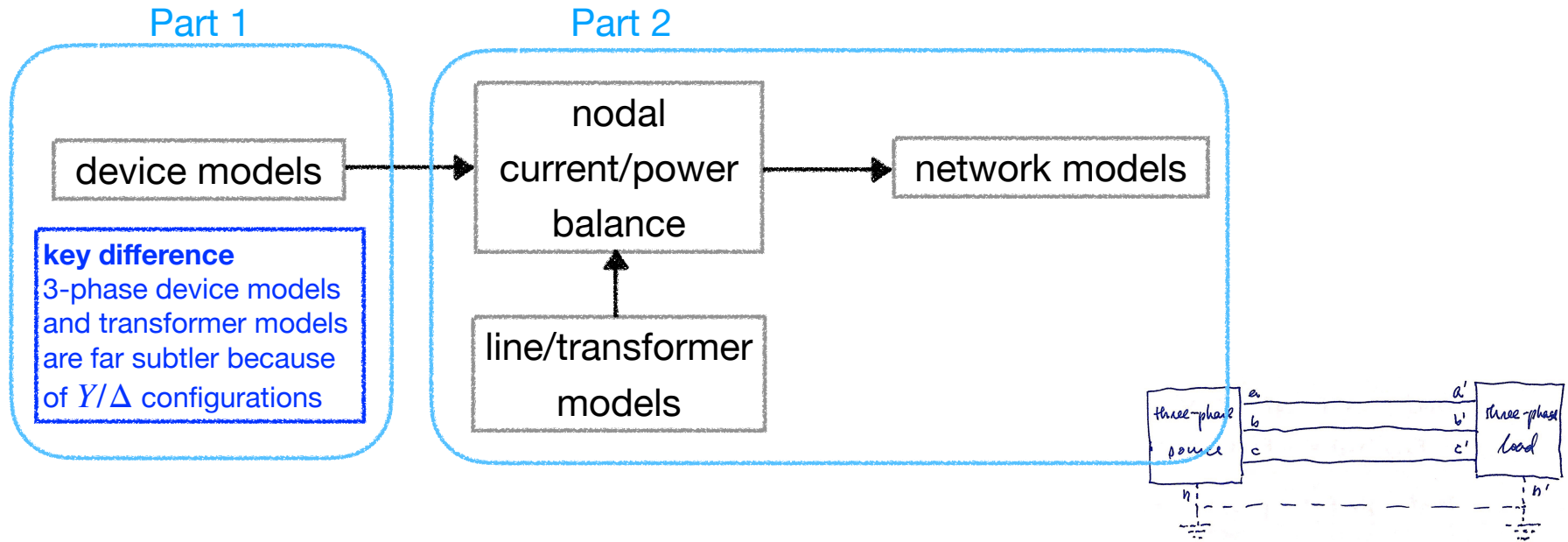
3-phase Power Flow Model

Steven Low Caltech

IREP, July 2022



Overview



single-phase or 3-phase



Key question

How to derive **external models** of 3-phase devices

1. Voltage/current/power sources, impedances (1-phase device: internal models)
2. ... in Y/Δ configurations (conversion rules: int \rightarrow ext)
3. ... with or without neutral lines, grounded or ungrounded, zero or nonzero grounding impedances

Propose a simple and unified method to derive external models

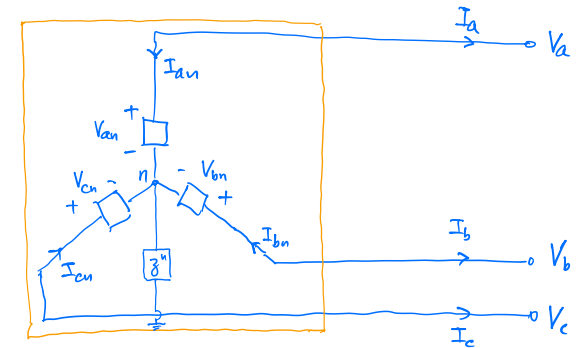


Internal variables

Y configuration

Internal voltage, current, power across **single-phase** devices:

$$V^Y := \begin{bmatrix} V^{an} \\ V^{bn} \\ V^{cn} \end{bmatrix}, \quad I^Y := \begin{bmatrix} I^{an} \\ I^{bn} \\ I^{cn} \end{bmatrix}, \quad S^Y := \begin{bmatrix} S^{an} \\ S^{bn} \\ S^{cn} \end{bmatrix} := \begin{bmatrix} V^{an} \bar{I}^{an} \\ V^{bn} \bar{I}^{bn} \\ V^{cn} \bar{I}^{cn} \end{bmatrix}$$



neutral voltage (wrt common reference pt) $V^n \in \mathbb{C}$

neutral current (away from neutral) $I^n \in \mathbb{C}$

Device may or may not be grounded, and neutral impedance z^n may or may not be zero

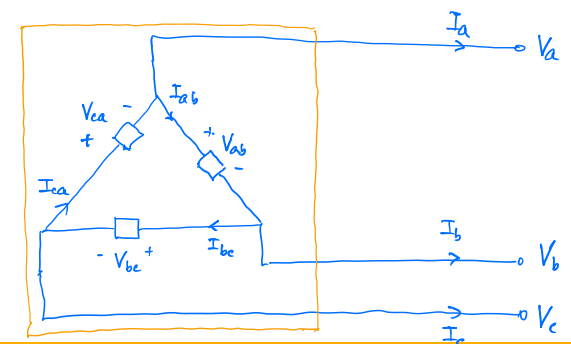


Internal variables

Δ configuration

Internal voltage, current, power across [single-phase](#) devices:

$$V^\Delta := \begin{bmatrix} V^{ab} \\ V^{bc} \\ V^{ca} \end{bmatrix}, I^\Delta := \begin{bmatrix} I^{ab} \\ I^{bc} \\ I^{ca} \end{bmatrix}, S^\Delta := \begin{bmatrix} S^{ab} \\ S^{bc} \\ S^{ca} \end{bmatrix} := \begin{bmatrix} V^{ab} \bar{I}^{ab} \\ V^{bc} \bar{I}^{bc} \\ V^{ca} \bar{I}^{ca} \end{bmatrix}$$



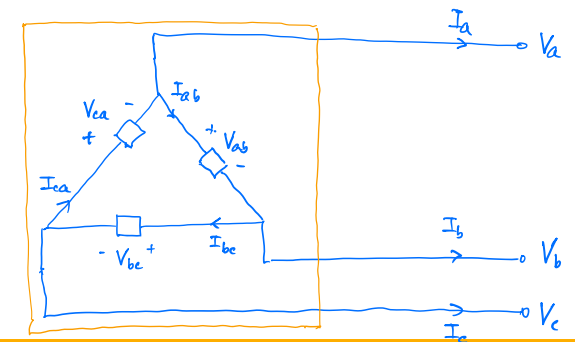
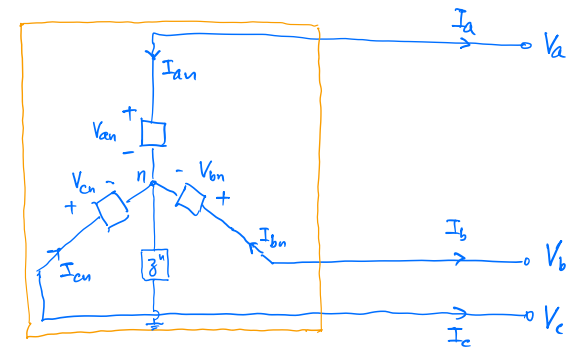


Terminal variables

Terminal voltage, current, power (for both Y and Δ) to reference:

$$V := \begin{bmatrix} V^a \\ V^b \\ V^c \end{bmatrix}, \quad I := \begin{bmatrix} I^a \\ I^b \\ I^c \end{bmatrix}, \quad s := \begin{bmatrix} s^a \\ s^b \\ s^c \end{bmatrix} := \begin{bmatrix} V^a \bar{I}^a \\ V^b \bar{I}^b \\ V^c \bar{I}^c \end{bmatrix}$$

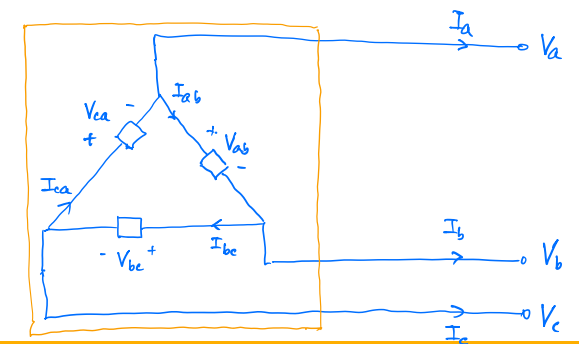
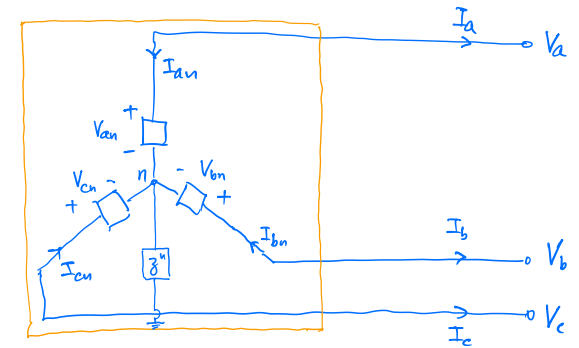
- V is with respect to an arbitrary common reference point, e.g. the ground
- I and s are in the direction **out** of the device





Internal vs external model

1. **External model** = Internal model + Conversion rule
 - External model: relation between (V, I, s)
 - Devices interact over network **only** through their terminal vars
2. **Internal model** : relation between $(V^{Y/\Delta}, I^{Y/\Delta}, s^{Y/\Delta})$
 - Independent of Y or Δ configuration
 - Depends only on behavior of single-phase devices
 - Voltage/current/power source, impedance
3. **Conversion rule** : converts between internal and terminal vars
 - Depends only on Y or Δ configuration
 - Independent of type of single-phase devices



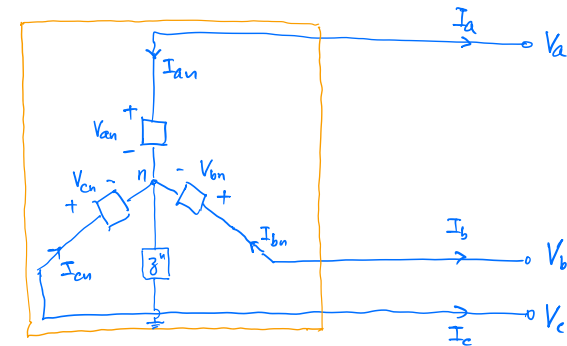


Conversion rule

Y configuration

Converts between internal and terminal variables

$$V = V^Y + V^n \mathbf{1}, \quad I = -I^Y, \quad s = - (s^Y + V^n \bar{I}^Y)$$



Device may or may not be grounded, and neutral impedance z^n may or may not be zero

Special case: if $V^n = 0$, then $V = V^Y$, $I = -I^Y$



Conversion rule

Δ configuration

Convert between **internal** vars and **external** vars

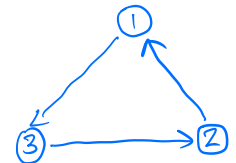
$$\begin{bmatrix} V_{ab} \\ V_{bc} \\ V_{ca} \end{bmatrix} = \underbrace{\begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & -1 \\ -1 & 0 & 1 \end{bmatrix}}_{\Gamma} \begin{bmatrix} V_a \\ V_b \\ V_c \end{bmatrix}, \quad \begin{bmatrix} I_a \\ I_b \\ I_c \end{bmatrix} = - \underbrace{\begin{bmatrix} 1 & 0 & -1 \\ -1 & 1 & 0 \\ 0 & -1 & 1 \end{bmatrix}}_{\Gamma^T} \begin{bmatrix} I_{ab} \\ I_{bc} \\ I_{ca} \end{bmatrix}$$

In vector form

$$\boxed{V^\Delta = \Gamma V, \quad I = -\Gamma^T I^\Delta}$$

↑ internal voltage ↑ terminal voltage ↑ terminal current ↑ internal current

Γ is incidence matrix of:





Conversion matrices

Fortescue matrix F

Spectral decomposition:

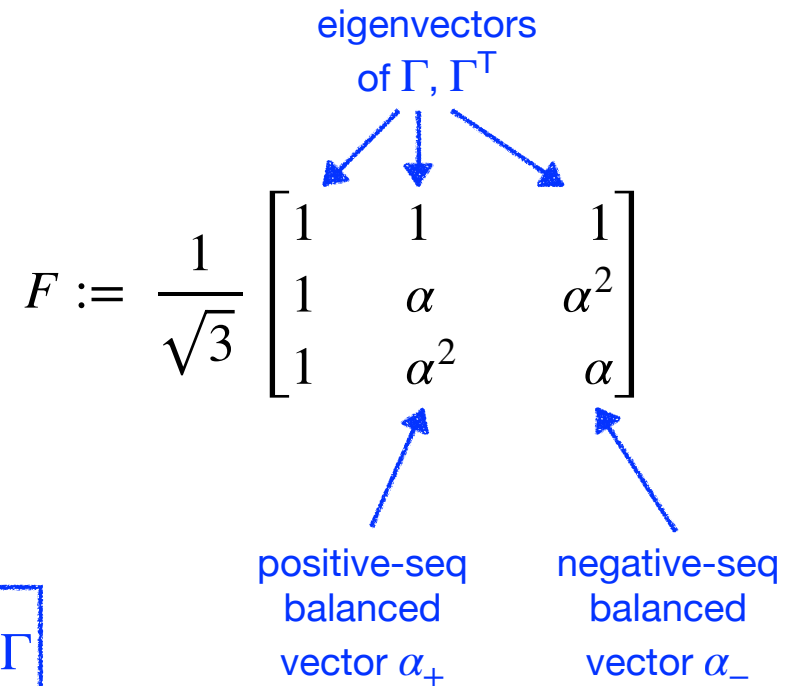
$$\Gamma = F\Lambda\bar{F}, \quad \Gamma^\top = \bar{F}\Lambda F$$

where

$$\Lambda := \begin{bmatrix} 0 & & \\ & 1 - \alpha & \\ & & 1 - \alpha^2 \end{bmatrix},$$

and $\alpha := e^{-i2\pi/3}$

Pseudo-inverses: $\Gamma^\dagger = \frac{1}{3}\Gamma^\top, \quad \Gamma^{\top\dagger} = \frac{1}{3}\Gamma$





Conversion rule

Δ configuration

1. Converts between internal and terminal voltages & currents

$$V^\Delta = \Gamma V, \quad I = -\Gamma^T I^\Delta$$

2. Given V^Δ : terminal voltage $V = \frac{1}{3} \Gamma^T V^\Delta + \gamma \mathbf{1}, \quad \gamma \in \mathbb{C}$

• $\gamma := \frac{1}{3} \mathbf{1}^T V$: zero-sequence terminal voltage (fixed by reference voltage)

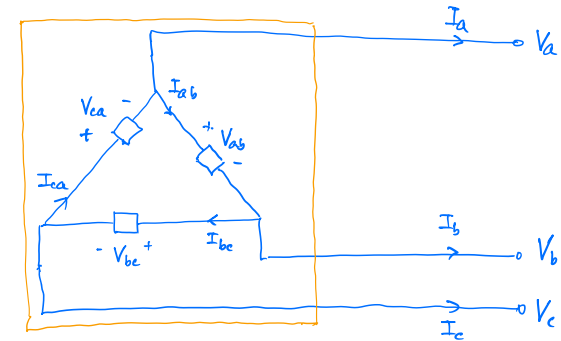
3. Given I : internal current $I^\Delta = -\frac{1}{3} \Gamma I + \beta \mathbf{1}, \quad \beta \in \mathbb{C}$

• $\beta := \frac{1}{3} \mathbf{1}^T I^\Delta$: zero-sequence internal current (does not affect terminal current)

4. Relation between s and s^Δ through (V, I^Δ) :

$$s = -\text{diag}(VI^{\Delta H}\Gamma), \quad s^\Delta = \text{diag}(\Gamma VI^{\Delta H})$$

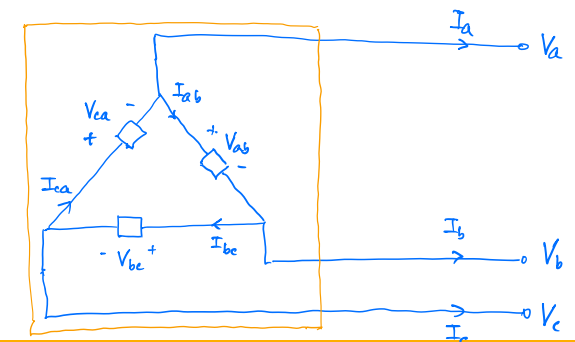
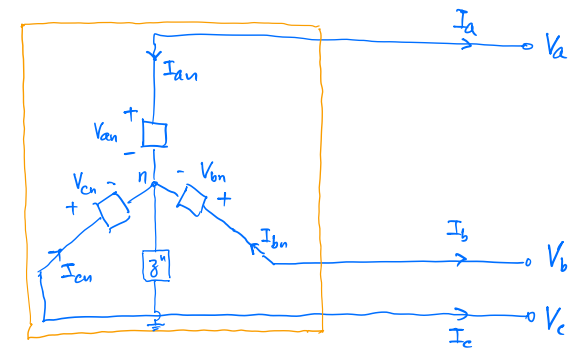
(no direct relation between s and s^Δ)





3-phase device models

1. **External model** = Internal model + Conversion rule
 - External model: relation between (V, I, s)
 - Internal model: relation between $(V^{Y/\Delta}, I^{Y/\Delta}, s^{Y/\Delta})$
2. Both internal and external models depend on device type
 - Voltage source
 - Current source
 - Power source
 - Impedance
3. ... in Y and Δ configurations





Voltage source (E^Δ, z^Δ) : Δ configuration

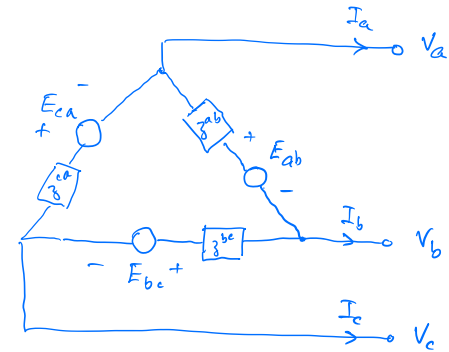
External model

1. Internal model

$$V^\Delta = E^\Delta + z^\Delta I^\Delta \quad \text{independent of } Y/\Delta \text{ config}$$

2. Conversion rule for Δ configuration

$$V^\Delta = \Gamma V, \quad I = -\Gamma^T I^\Delta$$





Voltage source (E^Δ, z^Δ) : Δ configuration

External model

1. Internal model

$$V^\Delta = E^\Delta + z^\Delta I^\Delta \quad \text{independent of } Y/\Delta \text{ config}$$

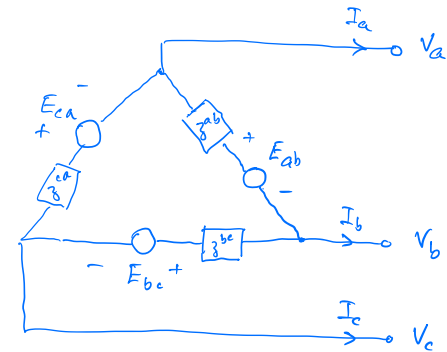
2. Conversion rule for Δ configuration

$$V^\Delta = \Gamma V, \quad I = -\Gamma^\top I^\Delta$$

3. Two (asymmetric) relations between terminal vars (V, I)

- Given V , 1st relation uniquely determines I (hence (V^Δ, I^Δ) as well)
- Given I , 2nd relation determines V up to zero-sequence voltage γ

Asymmetry is because V contains more info (γ) than I does (which contains no info about zero-sequence current $\beta := \frac{1}{3} \mathbf{1}^\top I^\Delta$)





Voltage source (E^Δ, z^Δ) : Δ configuration

External model

1. Given V ,

$$I = (\Gamma^\top y^\Delta) E^\Delta - Y^\Delta V$$

$$Y^\Delta := \Gamma^\top y^\Delta \Gamma = \begin{bmatrix} y^{ab} + y^{ca} & -y^{ab} & -y^{ca} \\ -y^{ab} & y^{ab} + y^{bc} & -y^{bc} \\ -y^{ca} & -y^{bc} & y^{ca} + y^{bc} \end{bmatrix}, \quad y^\Delta := (z^\Delta)^{-1}$$



Voltage source (E^Δ, z^Δ) : Δ configuration

External model

1. Given V ,

$$I = (\Gamma^\top y^\Delta) E^\Delta - Y^\Delta V$$

$$Y^\Delta := \Gamma^\top y^\Delta \Gamma = \begin{bmatrix} y^{ab} + y^{ca} & -y^{ab} & -y^{ca} \\ -y^{ab} & y^{ab} + y^{bc} & -y^{bc} \\ -y^{ca} & -y^{bc} & y^{ca} + y^{bc} \end{bmatrix}, \quad y^\Delta := (z^\Delta)^{-1}$$

2. Given I with $1^\top I = 0$,

$$V = \hat{\Gamma} E^\Delta - Z^\Delta I + \gamma 1, \quad 1^\top I = 0$$

$$\hat{\Gamma} := \frac{1}{3} \Gamma^\top \left(\mathbb{1} - \frac{1}{\zeta} \tilde{z}^\Delta 1^\top \right), \quad Z^\Delta := \frac{1}{9} \Gamma^\top z^\Delta \left(\mathbb{1} - \frac{1}{\zeta} 1 \tilde{z}^{\Delta\top} \right) \Gamma$$



Voltage source (E^Δ, z^Δ) : Δ configuration

External model

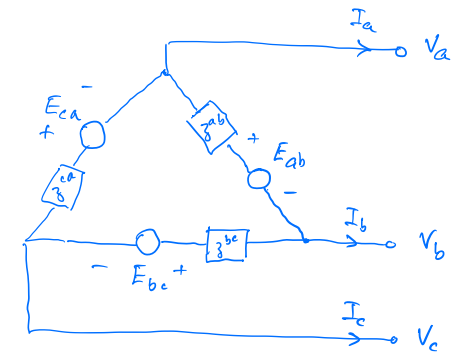
Comparison

Single-phase : $V = E - zI$

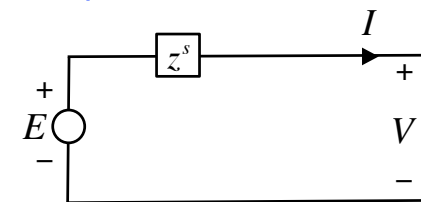
Three-phase : $V = \hat{\Gamma}E^\Delta - Z^\Delta I + \gamma 1, \quad 1^T I = 0$

rotated
internal voltage

voltage drop due to
equivalent impedance



1-phase device





Current source (J^Δ, y^Δ) : Δ configuration

External model

1. Internal model

$$I^\Delta = J^\Delta + y^\Delta V^\Delta$$

2. Conversion rule

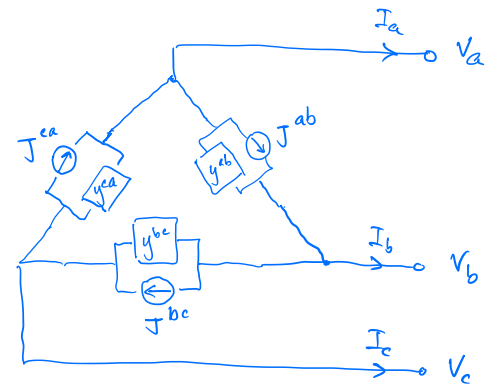
$$V^\Delta = \Gamma V, \quad I = -\Gamma^\top I^\Delta$$

3. \implies External model

$$I = -(\Gamma^\top J^\Delta + Y^\Delta V)$$

where (as before): $Y^\Delta := \Gamma^\top y^\Delta \Gamma =$

$$\begin{bmatrix} y^{ab} + y^{ca} & -y^{ab} & -y^{ca} \\ -y^{ab} & y^{ab} + y^{bc} & -y^{bc} \\ -y^{ca} & -y^{bc} & y^{ca} + y^{bc} \end{bmatrix}$$





Current source (J^Δ, y^Δ) : Δ configuration

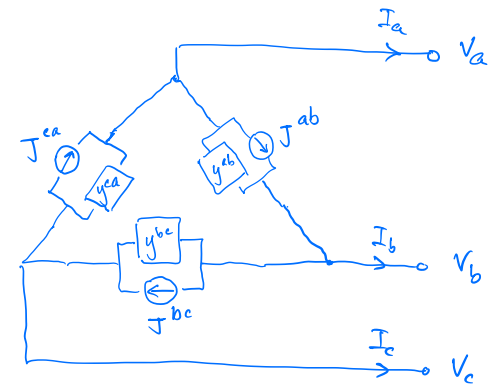
External model

4. Comparison

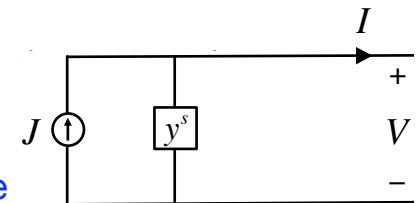
Single-phase : $I = J - yV$

Three-phase : $I = -\Gamma^T J^\Delta - Y^\Delta V$

$$Y^\Delta := \Gamma^T y^\Delta \Gamma$$



1-phase device



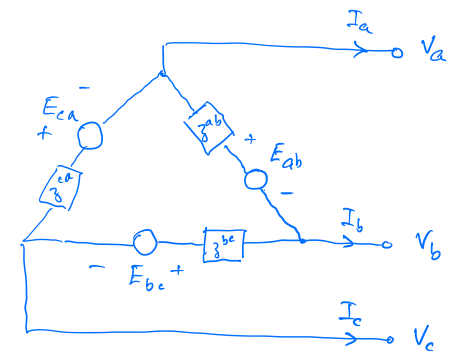
Note: directions of J are opposite



Voltage & current sources: comparison

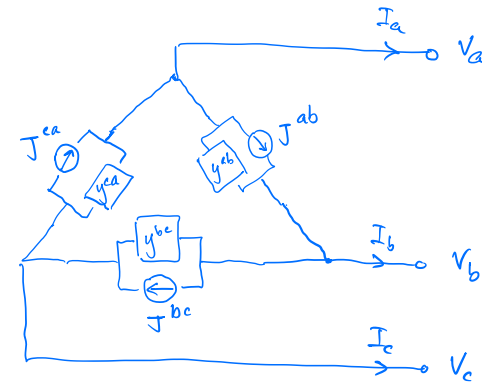
1. Voltage source specifies E^Δ which does not uniquely determine terminal voltage V

- $V = \hat{\Gamma}E^\Delta - Z^\Delta I + \gamma 1, \quad 1^\top I = 0$
- due to arbitrary zero-sequence voltage $\gamma := \frac{1}{3}1^\top V$



2. Current source specifies J^Δ which uniquely determines terminal current I

- $I = -(\Gamma^\top J^\Delta + Y^\Delta V)$
- J^Δ contains its zero-sequence current $\beta := \frac{1}{3}1^\top J^\Delta$





Impedance z^Δ : Δ configuration

External model

1. Internal model

$$V^\Delta = z^\Delta I^\Delta$$

2. Conversion rule

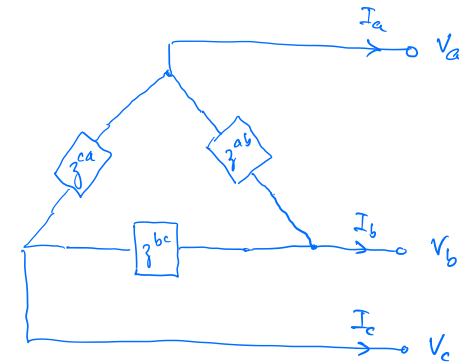
$$V^\Delta = \Gamma V, \quad I = -\Gamma^\top I^\Delta$$

3. \implies External model

Given V , $I = -Y^\Delta V := -(\Gamma^\top y^\Delta \Gamma) V$

Given I , $V = -Z^\Delta I + \gamma 1, \quad 1^\top I = 0$

$$Z^\Delta := \frac{1}{9} \Gamma^\top z^\Delta \left(\mathbb{1} - \frac{1}{\zeta} 1 \tilde{z}^{\Delta \top} \right) \Gamma$$



As for voltage source, the asymmetry is because V contains more info (γ) than I does



Impedance z^Δ : Δ configuration

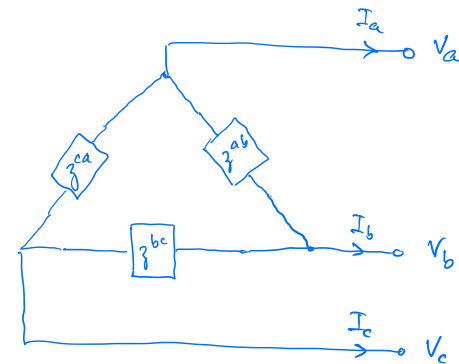
External model

4. Comparison

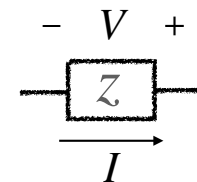
Single-phase : $V = -zI \in \mathbb{C}$

Three-phase : $V = -Z^\Delta I + \gamma 1, \quad 1^T I = 0$

↑
voltage drop due to
equivalent impedance



1-phase device





Power source σ^Δ : Δ configuration

External model

1. Internal model

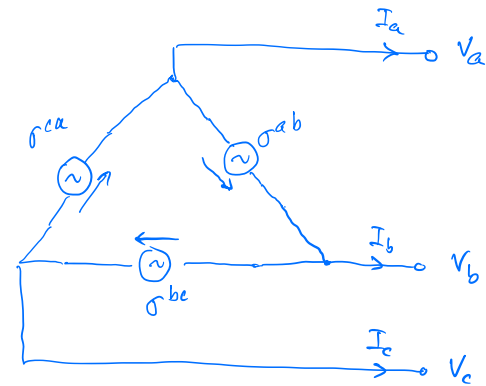
$$s^\Delta = \sigma^\Delta$$

2. Conversion rule

$$V^\Delta = \Gamma V, \quad I = -\Gamma^\top I^\Delta$$

3. \implies External model through (V, I^Δ)

$$s = -\text{diag}(VI^{\Delta H}\Gamma), \quad \sigma^\Delta = \text{diag}(\Gamma VI^{\Delta H})$$





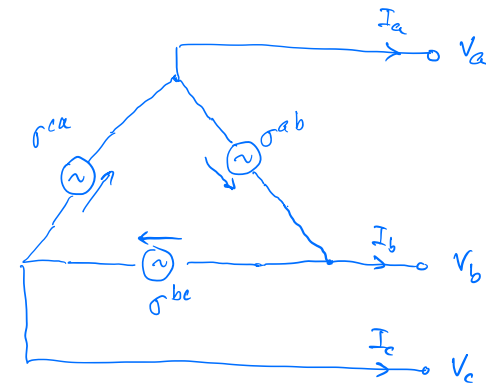
Power source σ^Δ : Δ configuration

External model

4. Comparison

Single-phase : $s = \sigma$

Three-phase : $s = -\text{diag}(VI^{\Delta H}\Gamma)$, $\sigma^\Delta = \text{diag}(\Gamma VI^{\Delta H})$

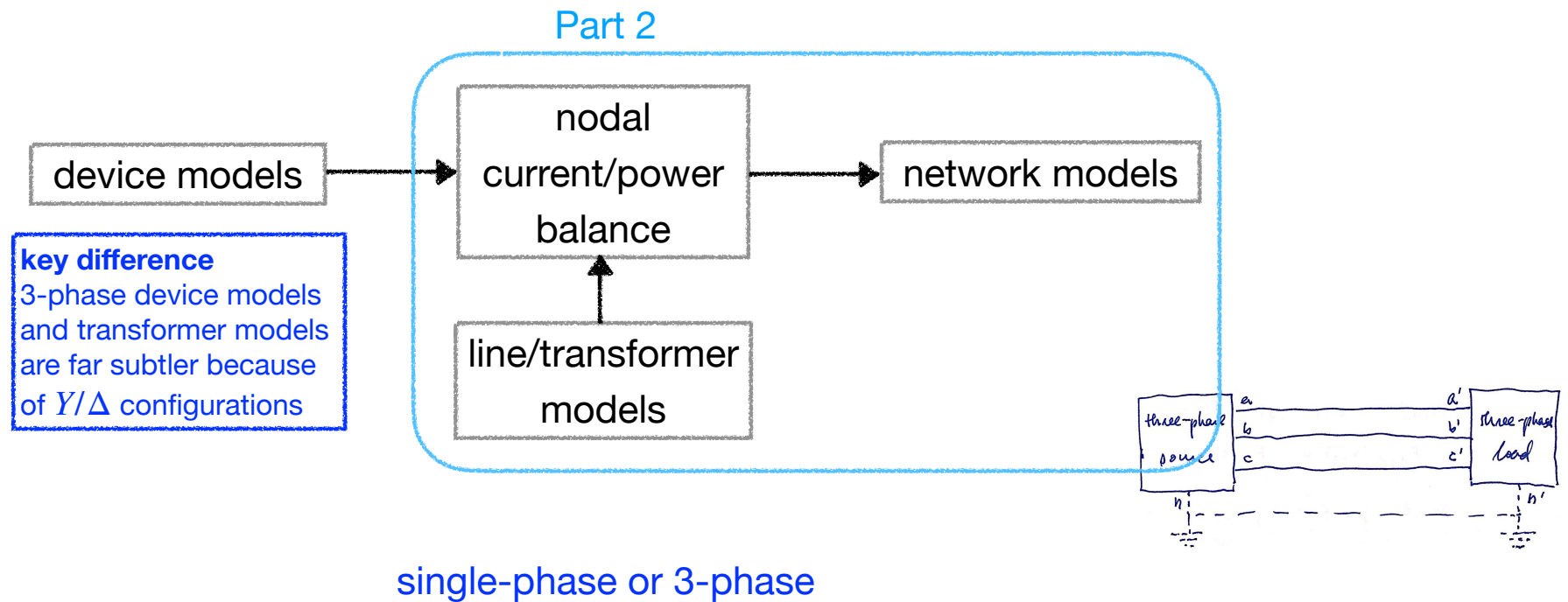


1-phase device





Overview



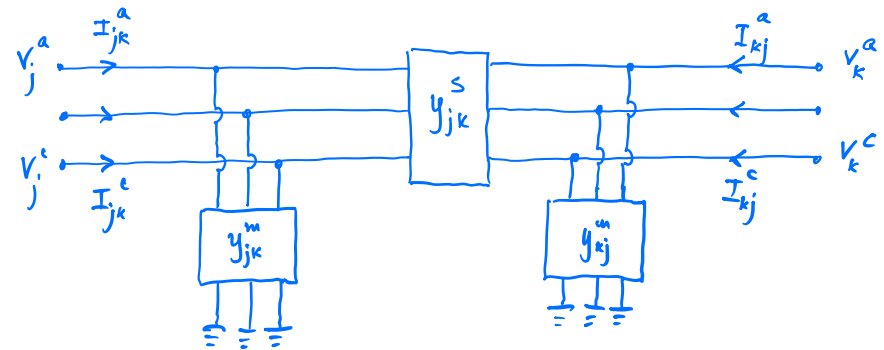


3-wire line model

With shunt admittances

Each line is characterized by

- Series admittance $y_{jk}^s := \left(z_{jk}^s\right)^{-1}$
- Shunt admittances $\left(y_{jk}^m, y_{kj}^m\right)$



Terminal voltages $\left(V_j, V_k\right)$ and terminal currents $\left(I_{jk}, I_{kj}\right)$ satisfy

$$I_{jk} = y_{jk}^s \left(V_j - V_k\right) + y_{jk}^m V_j$$

$$I_{kj} = y_{jk}^s \left(V_k - V_j\right) + y_{kj}^m V_k$$

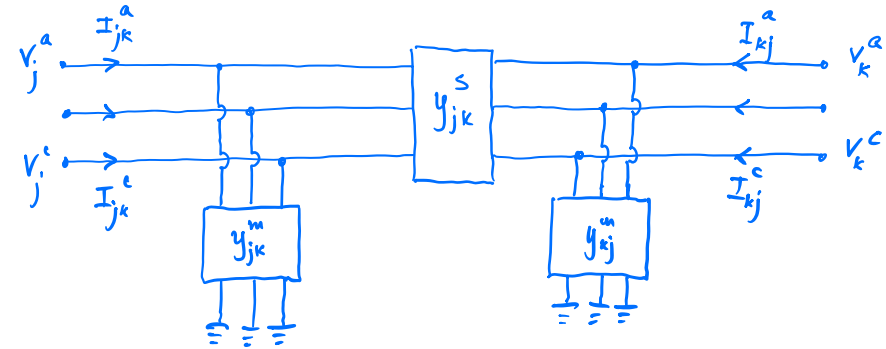


3-wire line model

With shunt admittances

Each line is characterized by

- Series admittance $y_{jk}^s := (z_{jk}^s)^{-1}$
- Shunt admittances (y_{jk}^m, y_{kj}^m)



Terminal voltages (V_j, V_k) and **terminal** power (S_{jk}, S_{kj}) satisfy

$$S_{jk} := V_j (I_{jk})^H = V_j (V_j - V_k)^H (y_{jk}^s)^H + V_j V_j^H (y_{jk}^m)^H$$

$$S_{kj} := V_k (I_{kj})^H = V_k (V_k - V_j)^H (y_{jk}^s)^H + V_k V_k^H (y_{kj}^m)^H$$



Network equation

Nodal current balance

3-phase sending-end currents:

$$I_{jk} = y_{jk}^s (V_j - V_k) + y_{jk}^m V_j, \quad I_{kj} = y_{jk}^s (V_k - V_j) + y_{kj}^m V_k$$

Series and shunt admittances

- 1-phase : scalars
- 3-phase : 3×3 (3-wire) or 4×4 (4-wire) matrices



Network equation

Nodal current balance

Series and shunt admittances

- 1-phase : scalars
- 3-phase : 3×3 (3-wire) or 4×4 (4-wire) matrices

3-phase sending-end currents:

$$I_{jk} = y_{jk}^s (V_j - V_k) + y_{jk}^m V_j, \quad I_{kj} = y_{jk}^s (V_k - V_j) + y_{kj}^m V_k$$

Nodal current balance:

$$\begin{aligned} I_j &= \sum_{k:j \sim k} I_{jk} = \sum_{k:j \sim k} y_{jk}^s (V_j - V_k) + \left(\sum_{k:j \sim k} y_{jk}^m \right) V_j \\ &= \left(\left(\sum_{k:j \sim k} y_{jk}^s \right) + y_{jj}^m \right) V_j - \sum_{k:j \sim k} y_{jk}^s V_k \end{aligned} \quad y_{jj}^m := \sum_{k:j \sim k} y_{jk}^m$$



Network equation

Nodal current balance

In terms of $3(N + 1) \times 3(N + 1)$ admittance matrix Y

$$I = YV \quad 3(N + 1) \text{ vector}$$

where

$$Y_{jj} := \sum_{k:j \sim k} y_{jk}^s + y_{jj}^m \quad 3 \times 3 \text{ matrices}$$

$$Y_{jk} := -y_{jk}^s \quad 3 \times 3 \text{ matrices}$$

$$y_{jj}^m := \sum_{k:j \sim k} y_{jk}^m$$

Y is complex (block-) symmetric [if network contains no 3-phase transformers in ΔY nor $Y\Delta$ config]

It is admittance matrix of single-phase equivalent



Network equation

Nodal power balance

Nodal power balance

$$s_j = \sum_{k:j \sim k} \text{diag} \left(V_j (V_j - V_k)^H (y_{jk}^s)^H + V_j V_j^H (y_{jk}^m)^H \right)$$

$$s_j = \text{diag} \left(V_j I_j^H \right)$$

generalizes single-phase:

$$s_j = \sum_{k:j \sim k} \left(|V_j|^2 - V_j V_k^H \right) (y_{jk}^s)^H + |V_j|^2 (y_{jj}^m)^H$$



Overall model

Device + network

1. **Network model** relates terminal vars (V, I, s)

- Nodal current balance (**linear**): $I = YV$

- Nodal power balance (**nonlinear**): $s_j = \sum_{k:j \sim k} \text{diag} \left(V_j (V_j - V_k)^H y_{jk}^{sH} + V_j V_j^H y_{jk}^{mH} \right)$

- Either can be used

2. **Device model** for each 3-phase device

- Internal model $\left(V_j^{Y/\Delta}, I_j^{Y/\Delta}, s_j^{Y/\Delta}, \gamma_j, \beta_j \right)$ + conversion rules

- External model $\left(V_j, I_j, s_j, \gamma_j, \beta_j \right)$ with internal parameters

- Either can be used

- Power source models are nonlinear; other devices are linear



General 3-phase analysis

Buses j	Specification
N_v^Y	$V_j^Y := E_j^Y, \gamma_j$
N_v^Δ	$V_j^\Delta := E_j^\Delta, \gamma_j, \beta_j,$
N_c^Y	$I_j^Y := J_j^Y, \gamma_j$
N_c^Δ	$I_j^\Delta := J_j^\Delta$
N_i^Y	z_j^Y, γ_j
N_i^Δ	z_j^Δ, β_j
N_p^Y	σ_j^Y, γ_j
N_p^Δ	$\sigma_j^\Delta, \gamma_j$

Variables at bus j :

- External vars : $(V_j, I_j, s_j), \gamma_j$
- Internal vars : $(V_j^{Y\Delta}, I_j^{Y/\Delta}, s_j^{Y/\Delta}), \beta_j$

Given: 3-phase devices & their **specifications**

- Voltage/current/power sources, impedances
- ... in Y/Δ configuration

Calculate: remaining variables

Solution:

- Write down **device+network** model
- Solve numerically



General 3-phase optimization

Buses j	Specification
N_v^Y	$V_j^Y := E_j^Y, \gamma_j$
N_v^Δ	$V_j^\Delta := E_j^\Delta, \gamma_j, \beta_j,$
N_c^Y	$I_j^Y := J_j^Y, \gamma_j$
N_c^Δ	$I_j^\Delta := J_j^\Delta$
N_i^Y	z_j^Y, γ_j
N_i^Δ	z_j^Δ, β_j
N_p^Y	σ_j^Y, γ_j
N_p^Δ	$\sigma_j^\Delta, \gamma_j$

Variables at bus j :

- External vars : $(V_j, I_j, s_j), \gamma_j$
- Internal vars : $(V_j^{Y\Delta}, I_j^{Y\Delta}, s_j^{Y\Delta}), \beta_j$

Given: 3-phase devices & uncontrollable quantities

- Voltage/current/power sources, impedances
- ... in Y/Δ configuration

Min: cost (controllable variables & state)

Solution:

- Write down device+network model
- Write down additional constraints
- Solve numerically



Unbalance 3-phase modeling

Power System Analysis

A Mathematical Approach

Steven H. Low

DRAFT available at: <http://netlab.caltech.edu/book/>

Corrections, questions, comments appreciated!