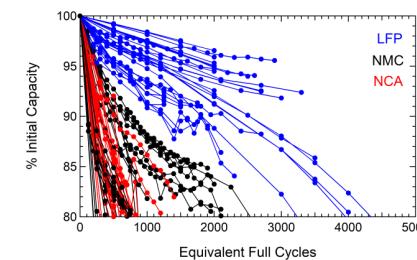
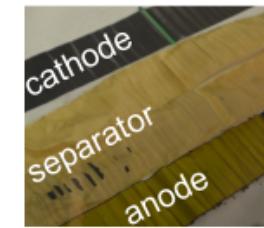
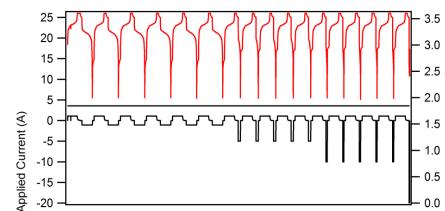


# A mechanism-based assessment of ‘knees’ in lithium-ion battery aging trajectories



*PRESENTED BY*

Yuliya Preger

IAPG Chemical Working Group Safety Panel

October 1, 2021

## 2 SNL grid energy storage program overview



Develop advanced energy storage technologies and systems, in collaboration with industry, academia, and government institutions, that will increase the reliability, performance, and competitiveness of electricity generation and transmission in the electric grid and in standalone systems.

### Materials

Advancing battery chemistries through development and commercialization

### Safety & Reliability

Testing, Analysis, Standards, Protocols

### Power Electronics

Reduce installed cost and footprint  
Improve control capability  
Increase reliability

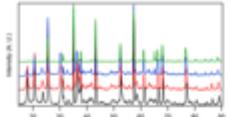
### Regulatory Outreach

Collaborating with States and other National Labs  
State Policy Analysis

### Demonstration Projects

Support, Analysis, Implementation, Monitoring

# Energy storage safety and reliability focus areas



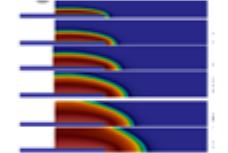
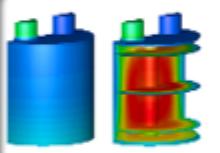
## Materials R&D

- Thermal stability and impact of aging on battery components
- Vent gas composition



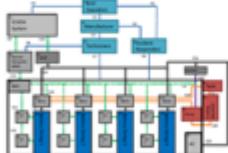
## Cell and Module Testing

- High precision cell cycling and degradation
- Electrical, thermal, mechanical abuse testing
- Failure propagation testing on batteries/systems



## Simulations and Modeling

- Multi-scale models for understanding thermal runaway
- Fire Dynamic Simulations to predict the size, scope, and consequences of battery fires



## System Level Design and Analysis

- Hazard analysis methods to avoid fire and explosion
- Predictive maintenance
- Power electronics control



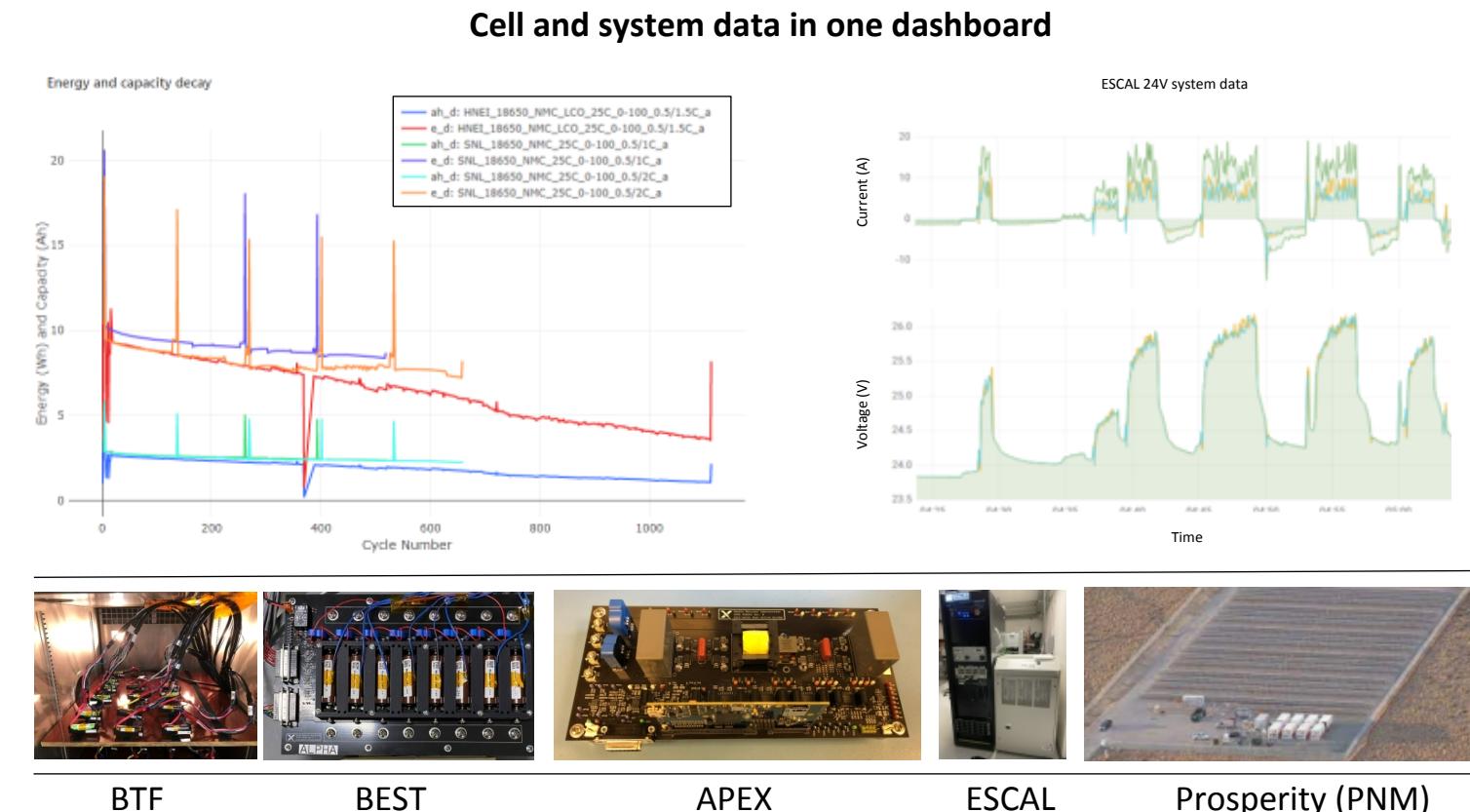
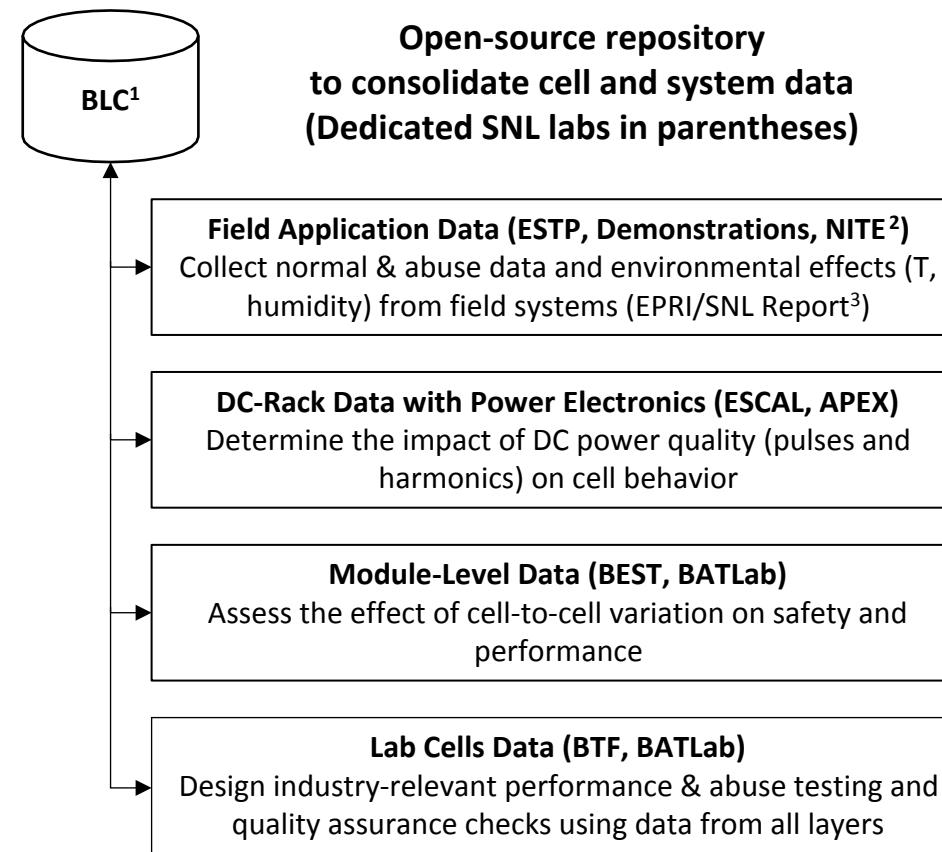
## Outreach, Codes, and Standards

- Energy storage safety working group
- IEEE battery management system standard
- EPRI data submission guidelines

# System operation and safety prediction: closing the data gap between cells and systems



We built labs and software to collect and consolidate data from all levels of battery operation for rapid technology iteration. The testing conditions are informed by environmental and system-level effects we have observed in field applications.



(1) V. De Angelis, Y. Preger, B. Chalamala "Battery Lifecycle Framework: A Flexible Repository and Visualization Tool for Battery Data from Materials Development to Field Implementation" ECSarXiv, 2021

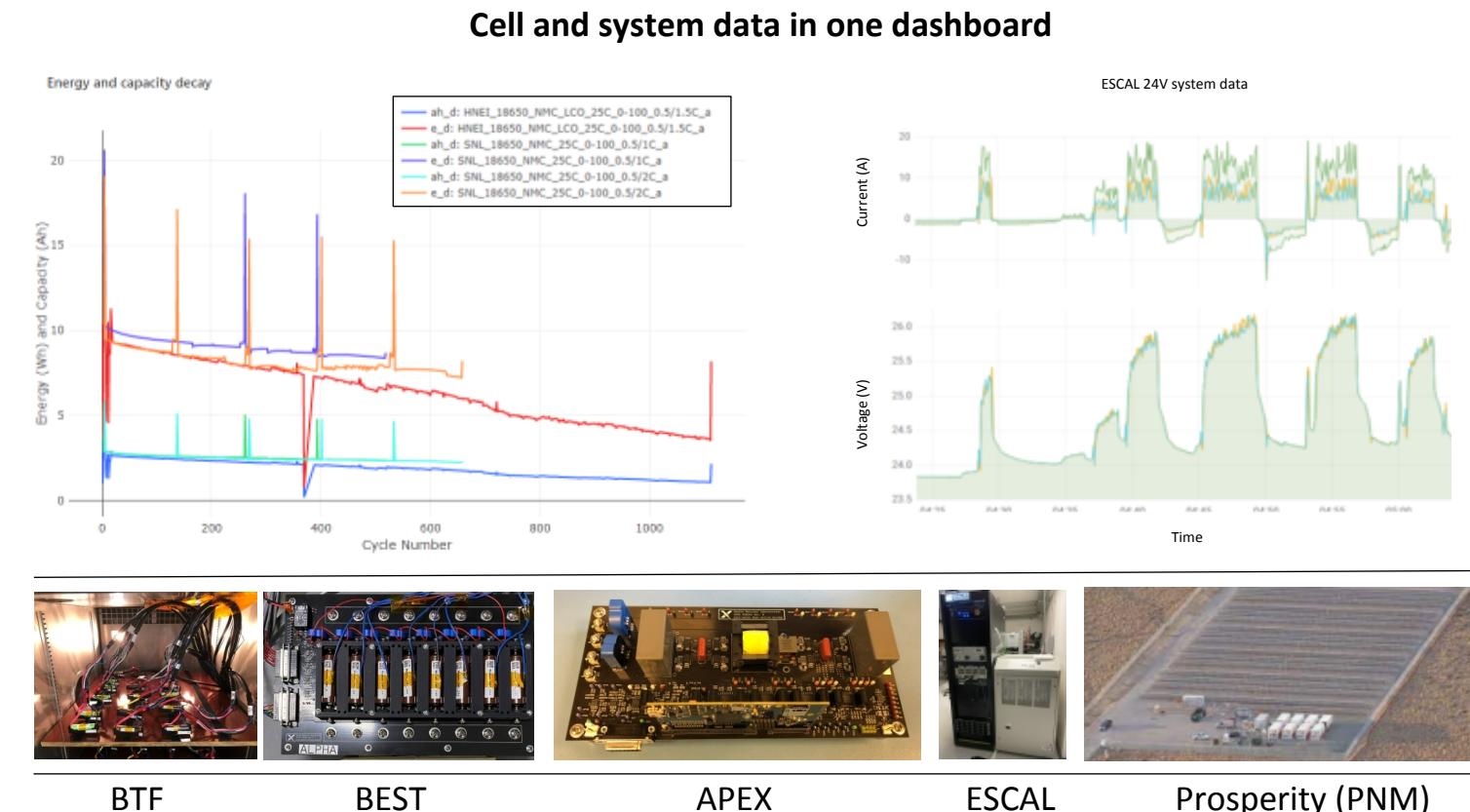
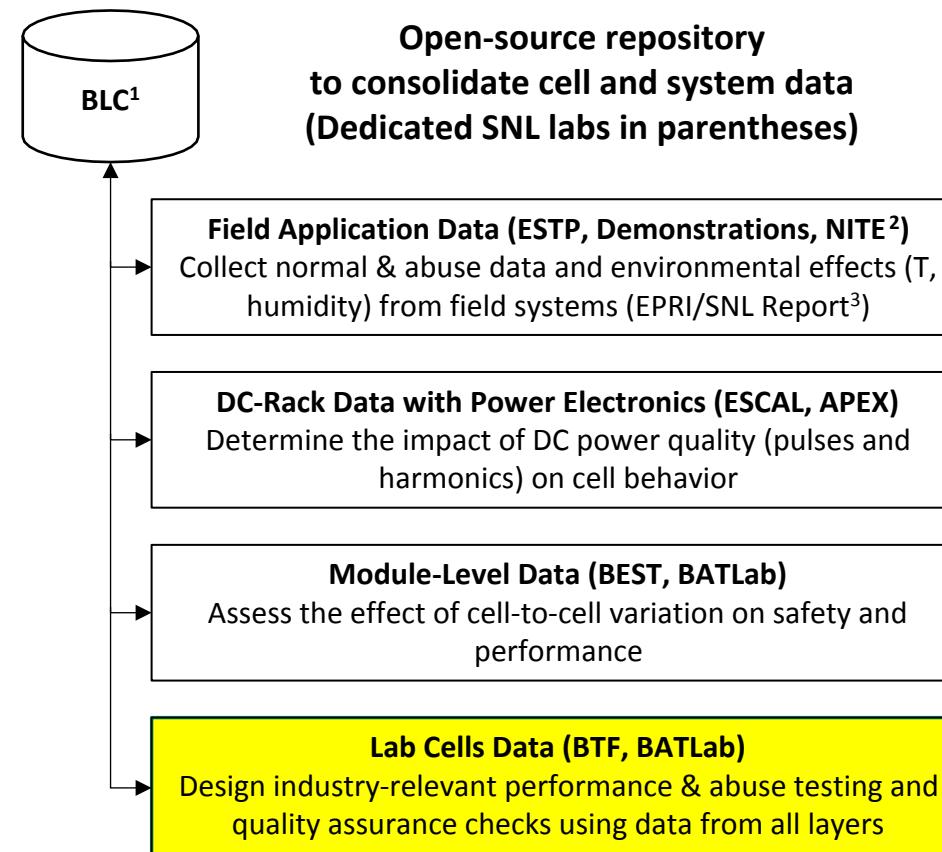
(2) B. Chalamala, D. Rosewater, Y. Preger, J. Lamb, R. Wittman, A. Kashiwakura "Ensuring Safety of Grid-scale Energy Storage Systems" IEEE Electrification, Vol. 9, No. 4, Dec. 2021

(3) D. Rosewater, Y. Preger, J. Mueller, S. Atcity, S. Willard, M. Smith, J. Thompson, D. Long, "Electrical Energy Storage Data Submission Guidelines, v 2" Electric Power Research Institute & SNL, 2021. 3002022119.

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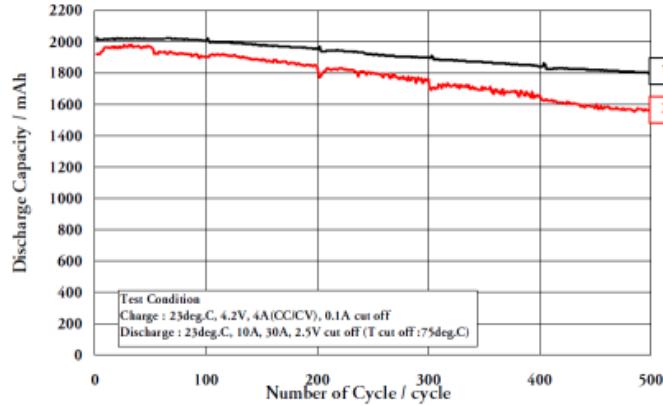
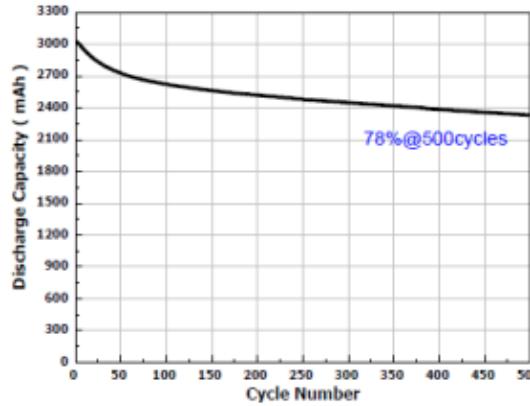
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# What is the remaining useful life (RUL) of a battery?

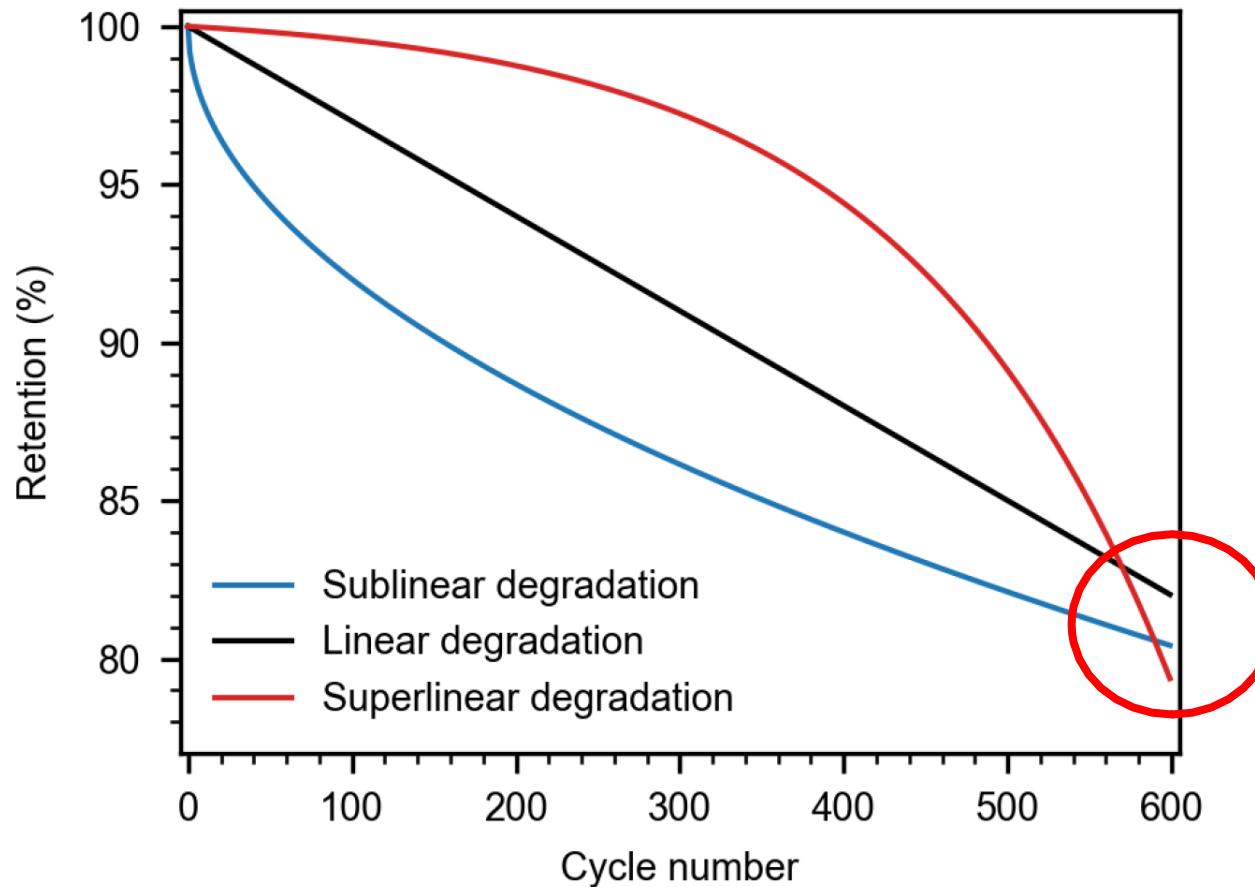


- 80% capacity is a common reference point in manufacturer spec sheets



- 80% capacity retention is a holdover from early EV days
  - USABC 1996: “EV batteries should be removed from automotive use when **current battery capacity is 80% of initial battery capacity** and **current battery power capability is 80% of initial battery power capability**”
  - At this time, EVs were primarily powered by Ni-based batteries
- Understanding RUL is critical for first life valuation and a deal-breaker for second life applications

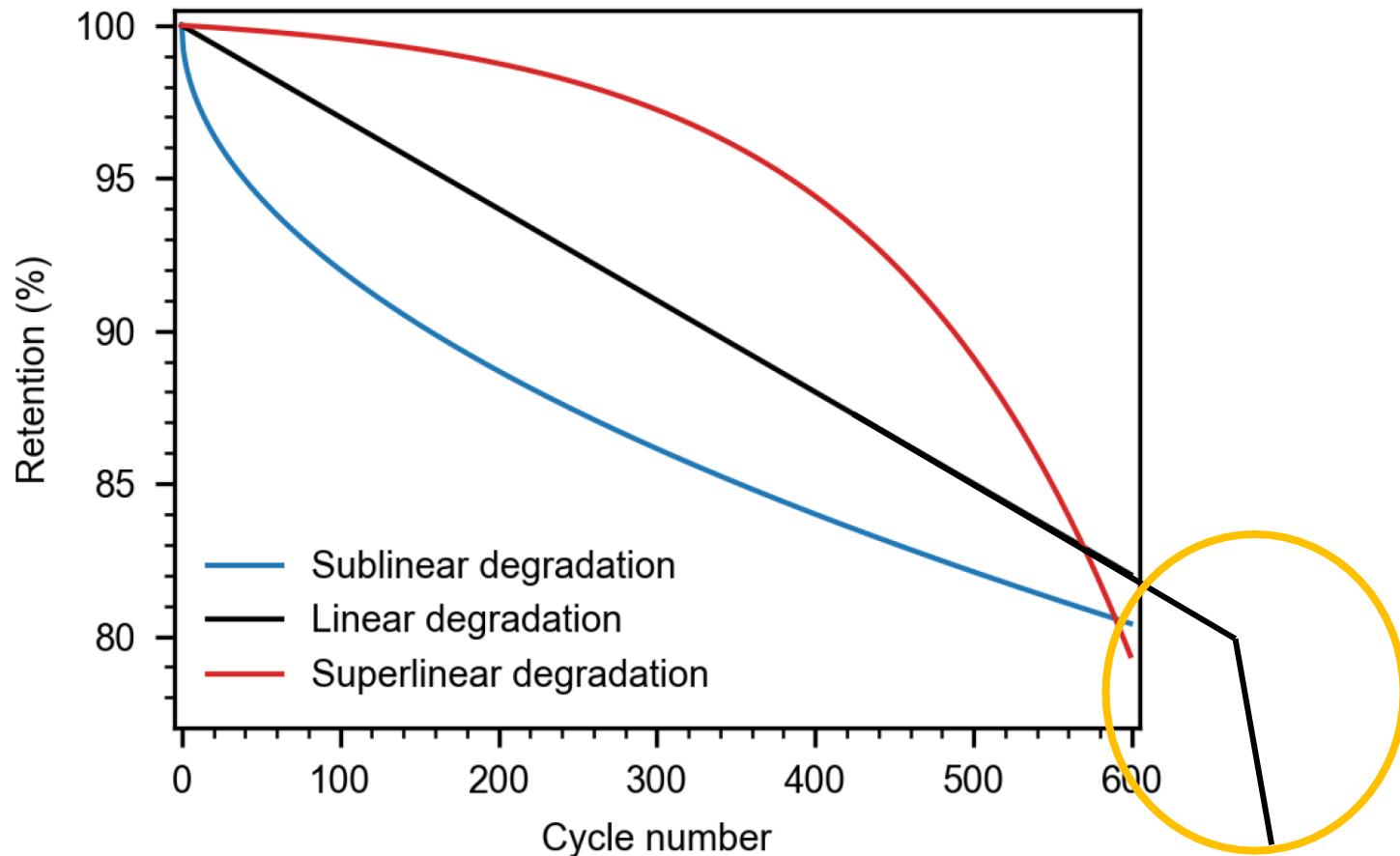
# Remaining useful life depends on the aging trajectory



Batteries with identical SOH have different RUL

What causes superlinear degradation (aka knee, rollover failure, nonlinear aging, two phase degradation, etc.) and how do you avoid it?

# Remaining useful life depends on the aging trajectory



How do you avoid this?

# Multi-institution team reviewing empirical causes and mechanisms of knee points



Stanford (Peter Attia)

Sandia (Yuliya Preger)

Carnegie Mellon University (Shashank Sripad, Alec Bills)

Hawai'i Natural Energy Institute (Matthieu Dubarry)

NREL (Paul Gasper)

University of Cincinnati (Abhishek Soni)

University of Michigan (Anna Stefanopoulou, Valentin Sulzer)

RWTH Aachen University (Philipp Dechent)

University of Edinburgh (Goncalo dos Reis, Richard Gilchrist)

University of Warwick (Ferran Brosa Planella)

University of Oxford (David Howey, Sam Greenbank)

A\*STAR (Edwin Khoo, Ouyang Liu)



# Project approach



- 1) Review all literature on knees: every experimental example + modeling explanation
- 2) Identify classes of degradation pathways
- 3) Determine how these pathways lend themselves to predictive capability

# “Perturbation” of Any Variable Can Induce Knees – Cell Design

11



Variable	Knee Acceleration	Reference
<b>Electrode loading</b>	Higher positive electrode loading	Ma et al. 2019
<b>Positive electrode coating</b>	Uncoated positive electrode	Ma et al. 2019
<b>Graphite type</b>	Natural graphite	Ma et al. 2019
<b>Additive package and concentration</b>	FEC consumed	Petibon et al. 2016
	FEC consumed	Jung et al. 2016
	Higher methyl acetate concentration	Ma et al. 2019
<b>Salt concentration</b>	Higher salt concentration	Aiken et al. 2020
	Lower salt concentration	Ma et al. 2019
	Higher salt concentration	Wang et al.

# “Perturbation” of Any Variable Can Induce Knees – Testing Conditions

12



Variable	Knee Acceleration	Reference
Charging rate	Higher charging rate	Lewerenz et al. 2017
	Higher charging rate	Petzl et al. 2015
	Higher charging rate	Burns et al. 2015
	Higher charging rate, constant voltage	Waldmann et al. 2015
	Higher charging rate	Schuster et al. 2015
	Higher charging rate	Severson et al. 2019
	Higher charging rate	Schindler et al. 2018
	Higher charging rate	Keil et al. 2019
Discharging rate	Lower discharging rate	Keil et al. 2016
	Lower discharging rate	Keil et al. 2019
	Lower discharging rate	Atalay et al. 2020
	Higher discharging rate	Omar et al. 2014
	No difference at 10-45 °C	Diao et al. 2019

# “Perturbation” of Any Variable Can Induce Knees – Testing Conditions

13



Variable	Knee Acceleration	Reference
Voltage limits	Higher SOC	Broussely et al. 2005
	Higher voltage	Aiken et al. 2020
	1) Higher DOD 2) Extreme midpoints	Ecker et al. 2014, Pfrang et al. 2018
	Higher DOD	Klett et al. 2014
	Higher DOD	Schuster et al. 2015
	1) Higher DOD 2) Higher midpoint SOC	Ma et al. 2019
	Higher DOD	Petzl et al. 2015
	Lower SOC	Zhu et al. 2021
Rests	Longer rest time	Keil et al. 2019
	Longer rest time	Ma et al. 2019
	Shorter rest time	Epding et al. 2019

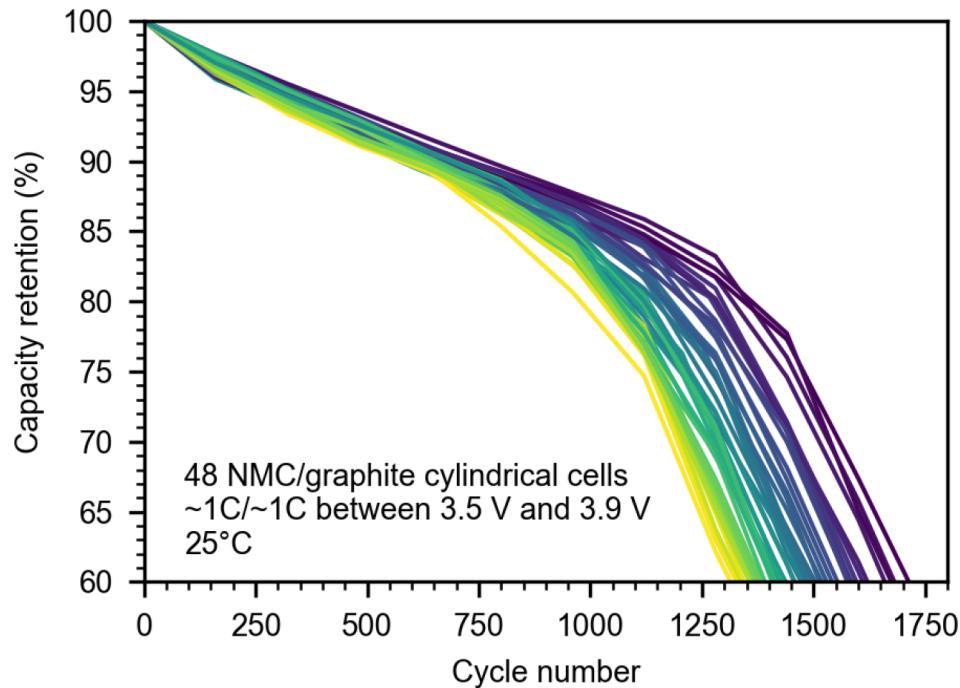
# “Perturbation” of Any Variable Can Induce Knees – Testing Conditions

14

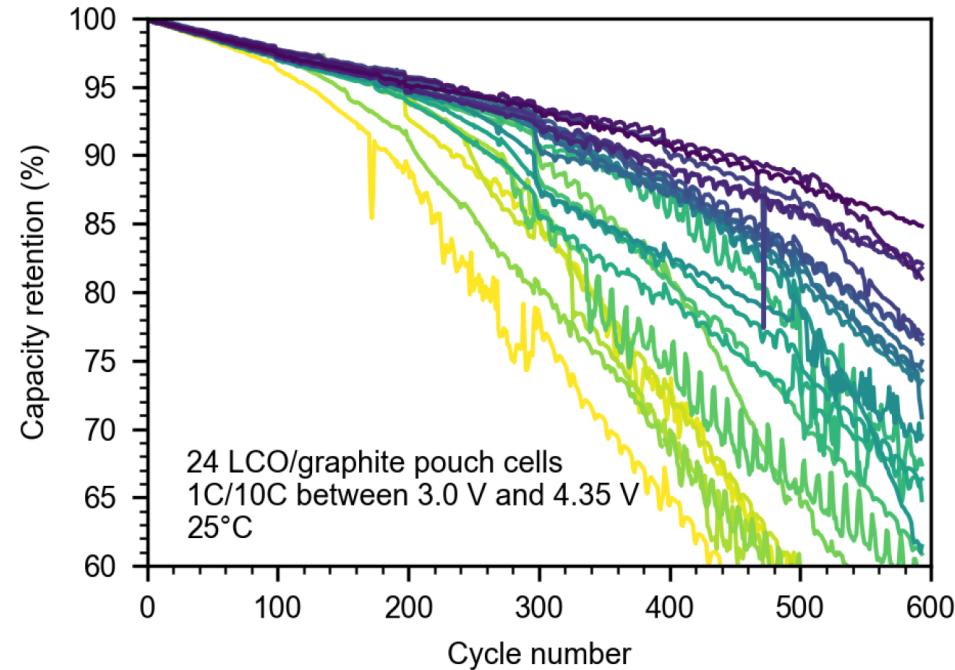


Variable	Knee Acceleration	Reference
Temperature	Temperature above and below 25 °C	Zhang et al. 2019
	Higher temperature	Broussely et al. 2005
	Temperature above and below 35 °C	Schuster et al. 2015
	Higher temperature	Safari et al. 2011
	Temperature above and below 25 °C	Waldmann et al. 2014
	Lower temperature	Coron et al. 2020
	Temperature below 25 °C	Waldmann et al. 2015
Pressure	More rigid bracing or zero bracing	Wunsch et al. 2019
	Higher stack pressure or zero pressure	Cannarella et al. 2014
	Heterogeneous compression	Bach et al. 2016

# Impact of cell to cell variability



Baumhofer et al. *J. Power Sources*, 2014, 247, 332.



Harris et al. *J. Power Sources*, 2017, 342, 589.

# General assessment of experimental studies



Knees are complex and occur under many conditions

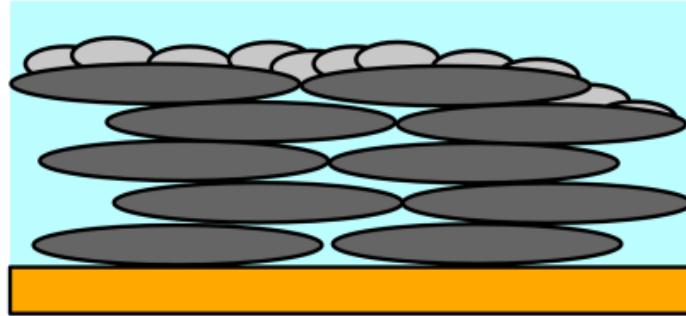
- Higher charging rate and wider DOD consistently accelerate knees
- Temperature/pressure have a 'sweet spot' outside of which knee is accelerated
- Discharge rate/rest time – it varies

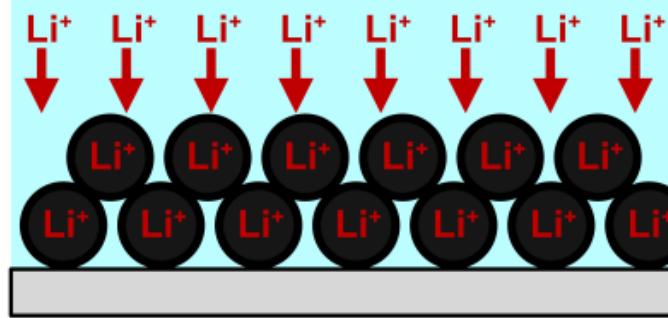
- Knees can occur during cycling within manufacturer specifications
- Knees observed as high as 90% remaining capacity and as low as 40%
- No specific range of values to avoid – specific value depends on the other variables

Next step: link all experimental observations to broader classes of degradation pathways

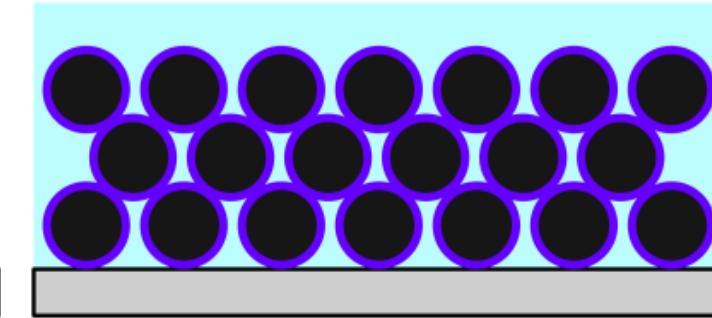
# Six pathways to knees



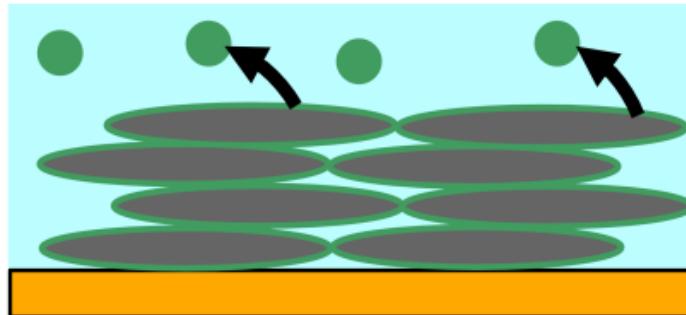
Lithium plating



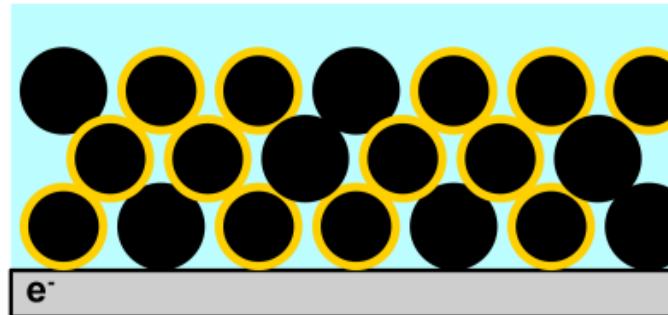
Electrode saturation



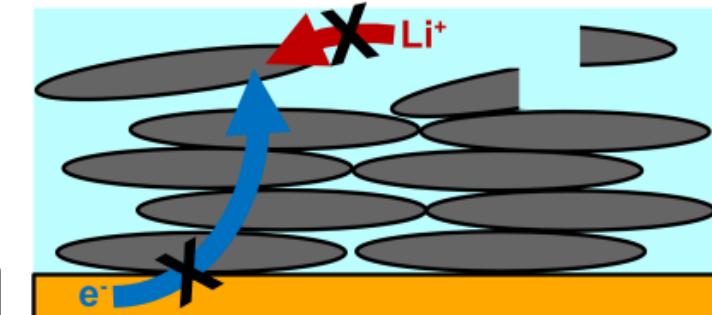
Resistance growth



Additive depletion

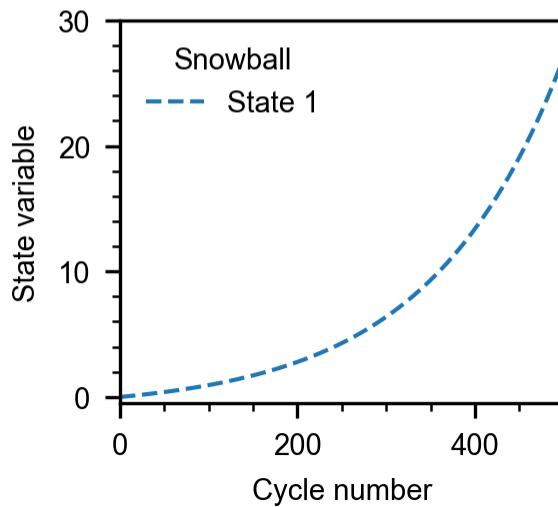
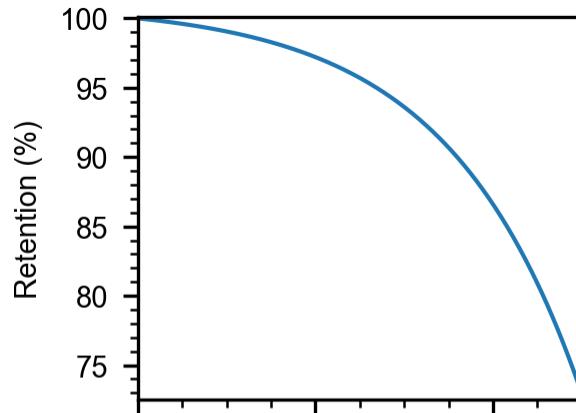


Percolation-limited connectivity



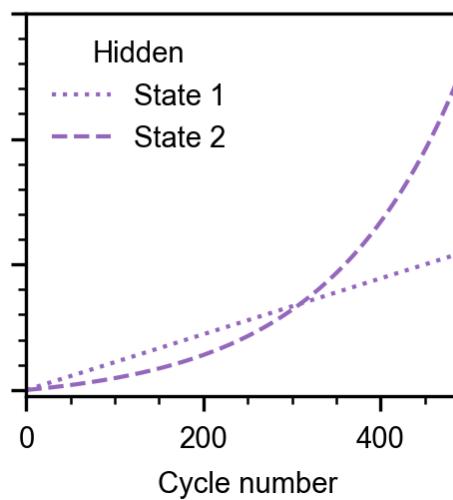
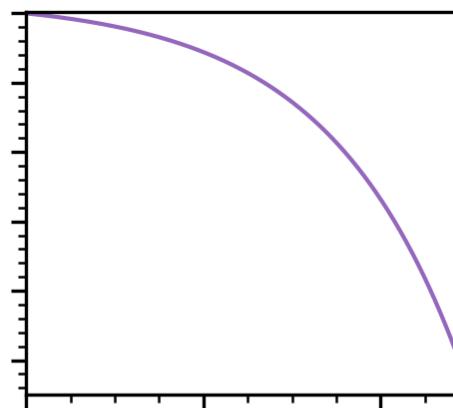
Mechanical deformation

# Defining internal state trajectories



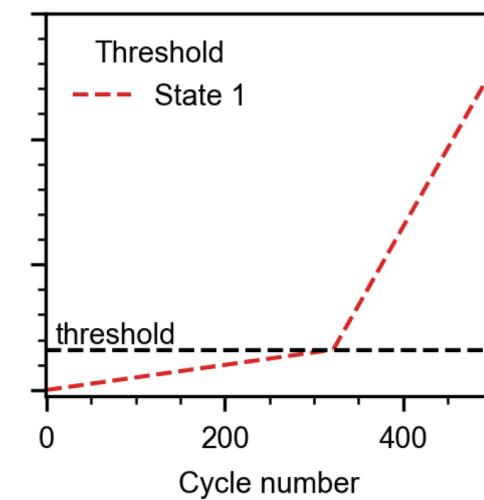
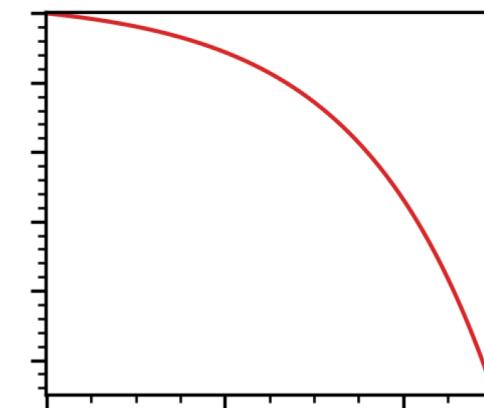
Ex. Mechanical deformation

*Extrapolation of exponential function  
difficult if there's noise*



Ex. Electrode saturation

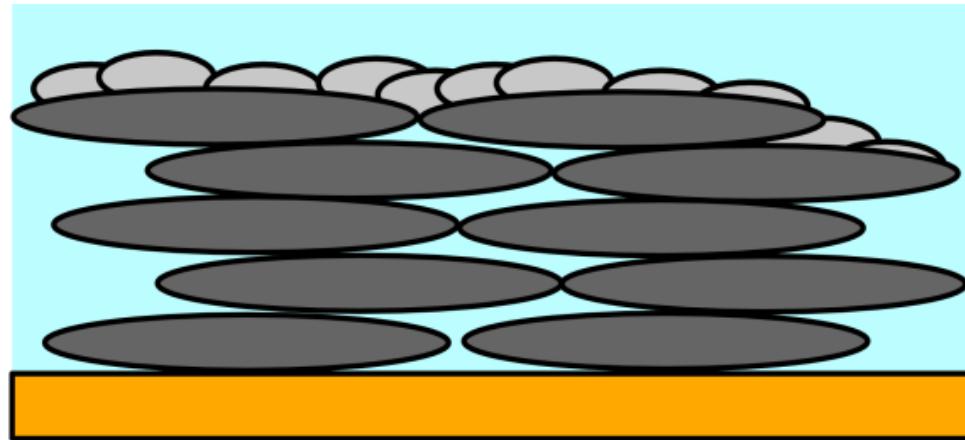
*Need to track two internal states*



Ex. Additive depletion

*Requires knowledge of internal  
state trajectory AND threshold*

# Pathway 1: Li plating



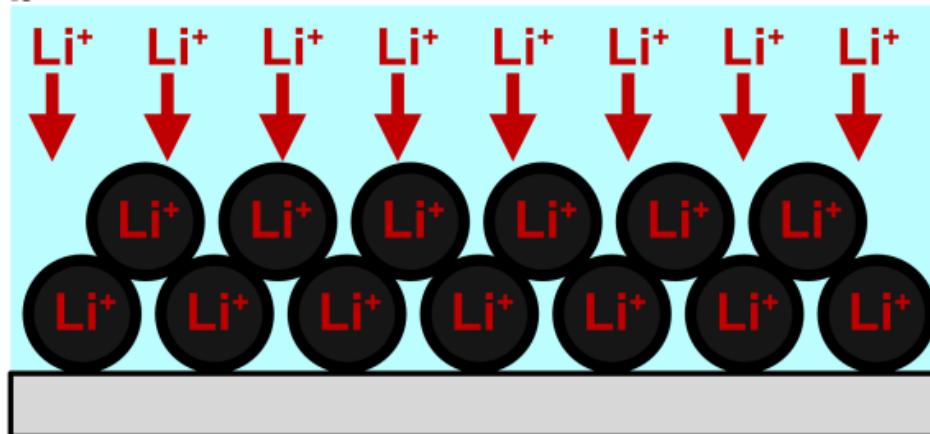
**What is it:** Li ions form metallic lithium on surface of negative electrode rather than intercalating

## Cause

- Rate independent: Loss of active material from delithiated negative electrode
- Rate dependent: High transport/reaction overpotentials cause local electrode potential to drop below  $\text{Li}/\text{Li}^+$ 
  - salt depletion, low temperature, high charge rate
  - mechanical compression reduces porosity
  - LAM increasing local current density
  - pore clogging from SEI build-up

**Trajectory:** snowball, hidden, threshold

## Pathway 2: Electrode saturation



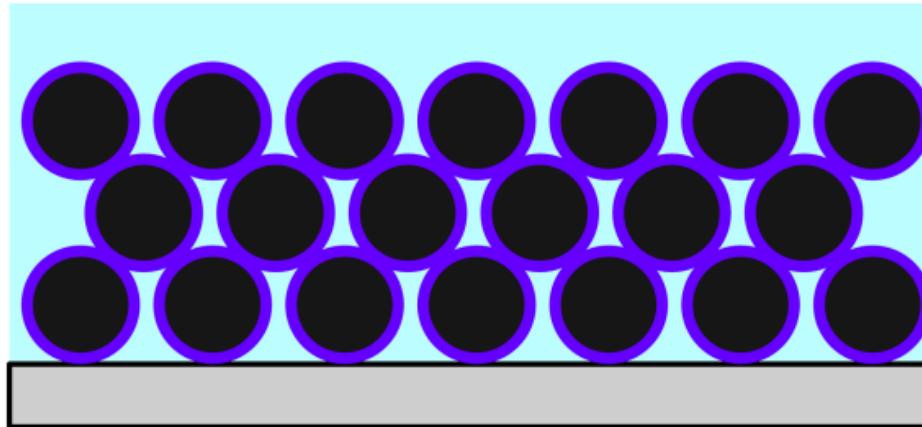
**What is it:** electrode “saturates” and reaches cutoff potential before all Li transferred

### Cause

- Rate of active material loss for one electrode outpaces lithium inventory loss
- Applicable to both electrodes (can be hidden for some time by oversized negative electrode)

**Trajectory:** threshold, hidden

## Pathway 3: Resistance growth



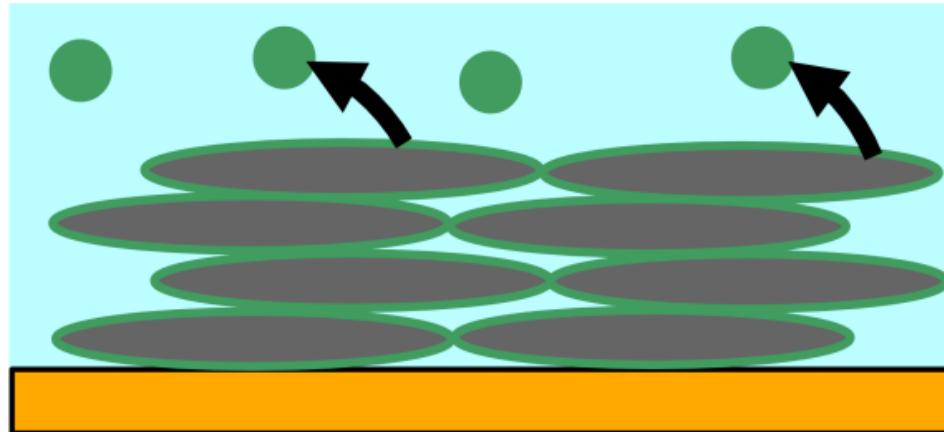
**What is it:** cell internal resistance increases, with additional overpotential causing cell to reach cutoff voltage quicker

### Cause

- Growth of side reaction products (electrolyte oxidation/reduction) on surface of electrode particles

**Trajectory:** threshold

## Pathway 4: Additive depletion



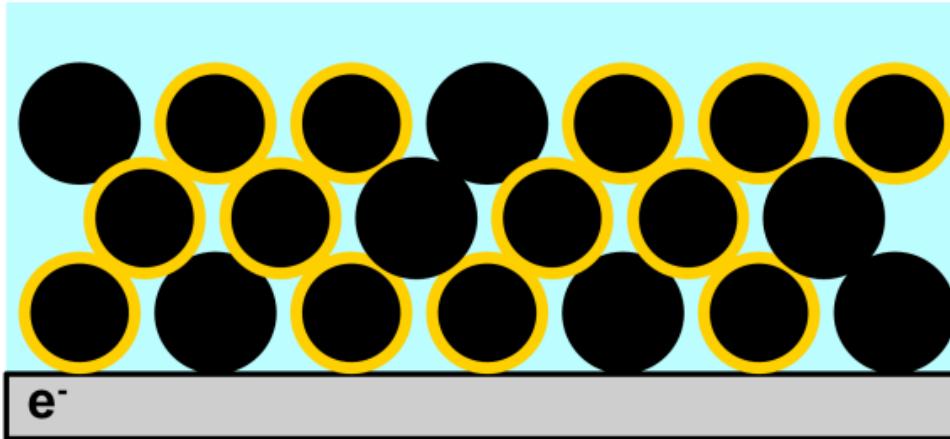
**What is it:** quantity of electrolyte additive in cell reduced over time

### Cause

- Additive consumed in side reactions
- More likely in commercial cells without excess electrolyte

**Trajectory:** threshold

## Pathway 5: Percolation-limited connectivity



**What is it:** electronic/ionic conductive network does not span the full electrode

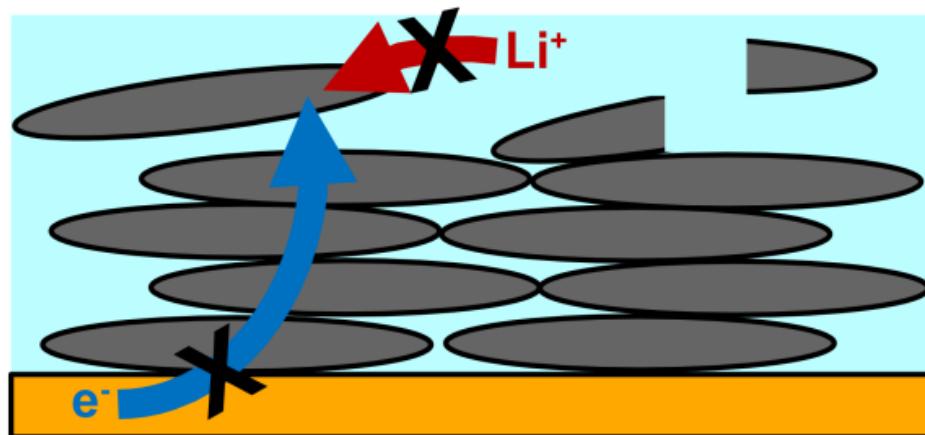
### Cause

- Electrode dry-out results in loss of ionic contact between active material and electrolyte, leading to loss of active material

**Trajectory:** threshold

# Pathway 6: Mechanical deformation

**What is it:** physical changes at the micro- and macro-scale due to mechanical processes



## Cause

- Micro-scale
  - particle cracking due to (de)intercalation stress
  - delamination
- Macro-scale
  - heterogeneous internal pressure distribution from cell components (e.g. tab)
  - jelly roll deformation causes loss of active material
  - uneven external pressure causes plating

**Trajectory:** snowball, threshold

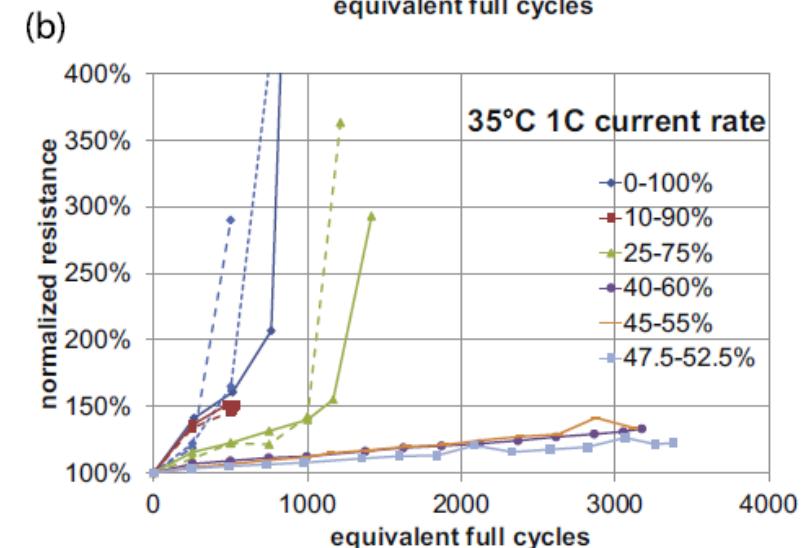
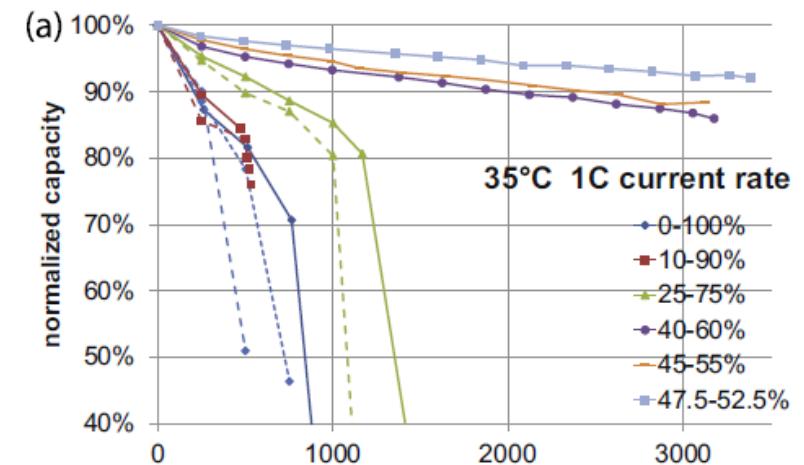
# Implications for modeling and prediction

- Goal of knee pathways evaluation: provide fundamental understanding of physics of knees to assess limits of today's models
- Lithium plating, electrode saturation, resistance growth
  - dependent on bulk internal states (LLI, LAM, etc.), so they are straightforward to detect and model via electrochemistry
- Additive depletion, percolation-limited connectivity, microscale mechanical deformation
  - involve subtle effects that are challenging to measure (e.g. porosity decrease, remaining additive amount)

# Value of resistance measurements for knee prediction in the field



- Capacity knee onset is nearly always correlated to onset of rapid resistance growth (elbow)
  - many intertwined degradation mechanisms
- Correlation is important because measuring capacity in the field can be hard, but measuring resistance at various SOCs is not



## Role of data-driven methods in knee prediction

- Methods are well-suited for knee pathways with bulk electrochemical signals
- Models trained on cycling data are poorly suited for knee pathways with signals that are challenging to measure via electrochemistry
- Datasets that span many pathways for various cell designs + use cases are needed for training generalizable models

# Conclusions



- Reviewed all examples of knees in the literature
  - Knees are complex and occur under many cycling + design conditions
  - No specific range of values to avoid – specific value depends on the other variables
- Identified classes of degradation pathways and underlying state trajectories
  - Li plating, electrode saturation, resistance growth, additive depletion, percolation-limited connectivity, mechanical deformation
  - Internal state trajectories (snowball, hidden, threshold) each pose unique challenges for monitoring
- Determined how pathways lend themselves to predictive capability
  - Need to consider which pathways have bulk electrochemical signals
  - Resistance is a useful value for field predictions

# Acknowledgments



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- S.S.: Carnegie Mellon University Presidential Fellowship

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