

Optimal Operation and Management of Energy Storage Systems Based on Real time Predictive Modeling and Adaptive Battery Management Techniques

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Performance Targets and Validation Plan

- **Integration of microcontroller with physics-based control models onto a large format cell and to demonstrate**
 - i) a 20% reduction in the weight of the cell
 - ii) a 50% reduction in the charging time for the cell without compromising the number of cycles
- **3 cells will be subject to cycling with no heat control in a chamber held at -17, 0 and 30°C for 3 test plans. Life of cells tested in case (B) will be better than (A), and ideally closer to base conditions (Test-0)**
 - Test 0: 30% to 85% SOC window
 - Test A: 0% to 100% SOC window
 - Test B: 0% to 100% SOC window with MPC based charging control

Cost / Benefit – Battery downsizing

- **Assumptions**

- 10 kWh useable system
- **Costs escalate at 2% per year**
- Discount rate of 5% per year is applied
- **10,000 vehicles produced per year**

- **Outcome**

- Present value of potential savings is **\$47M** for each vehicle model

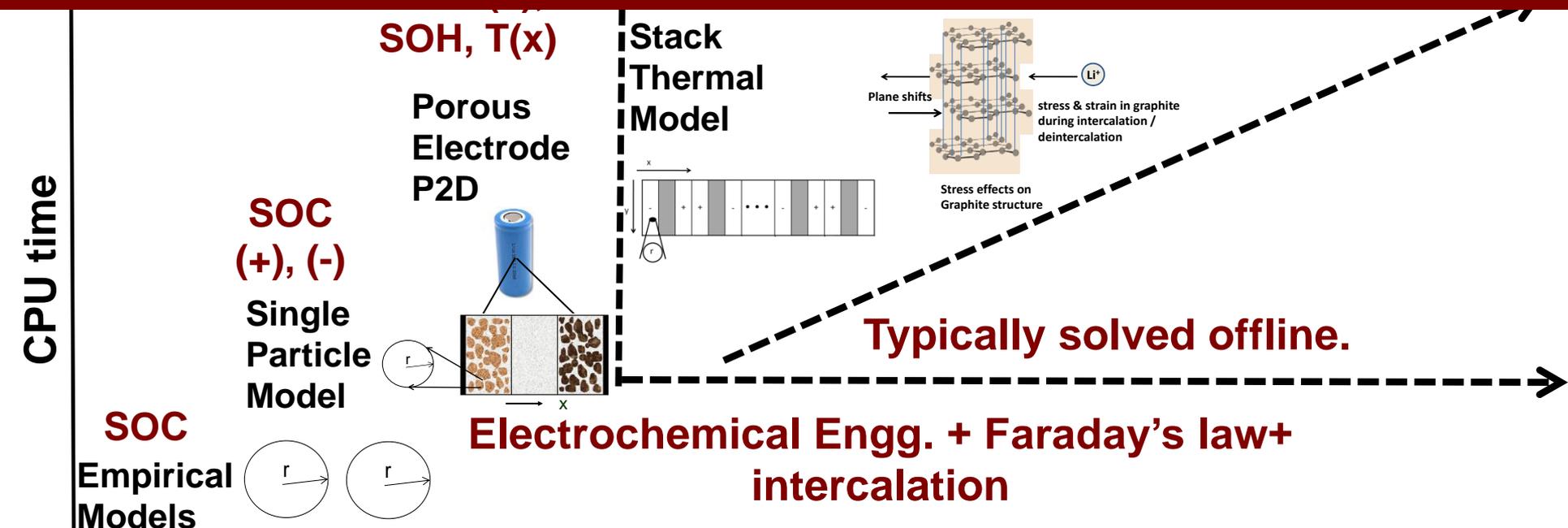
Useable Capacity	Today's Total Capacity	Total Capacity w/ AMPED model	Per vehicle battery cost savings	Per vehicle Additional BMS HW cost	Per vehicle development cost savings
1 kWh	1.7 kWh	1.11 kWh	\$111	\$35	-\$1.67
4 kWh	6.7 kWh	4.44 kWh	\$458	\$35	-\$1.67
6 kWh	10.0 kWh	6.67 kWh	\$690	\$35	-\$1.67
<u>10 kWh</u>	<u>16.7 kWh</u>	<u>11.1 kWh</u>	<u>\$1,152</u>	<u>\$35</u>	<u>-\$1.67</u>
20 kWh	33.3 kWh	22.2 kWh	\$2,310	\$35	-\$1.67

Models for batteries

MD, KMC, etc

Model Reformulation

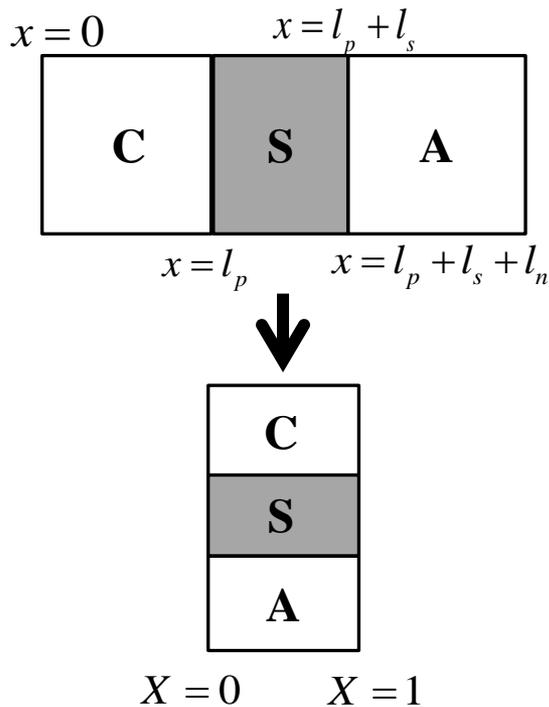
Reduces Computational Time



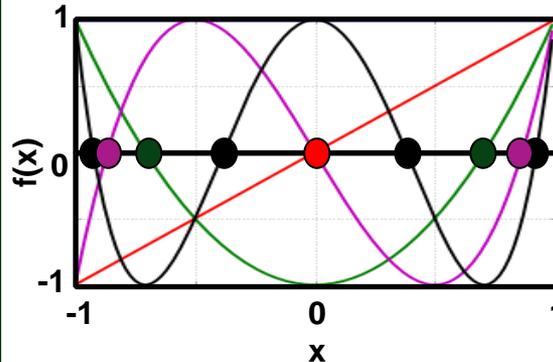
- Adding more physics provides more *fidelity* and *functionality* for the model and the BMS
- Comes at increased computational cost

Mathematical Reformulation

Coordinate Transformation



Orthogonal Collocation



Analytical Solutions

Original System
3000-10000 DAEs

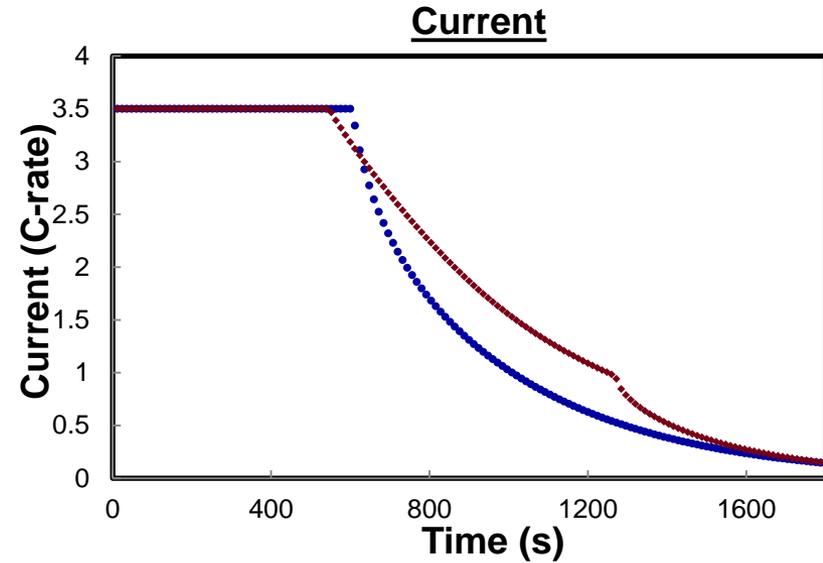
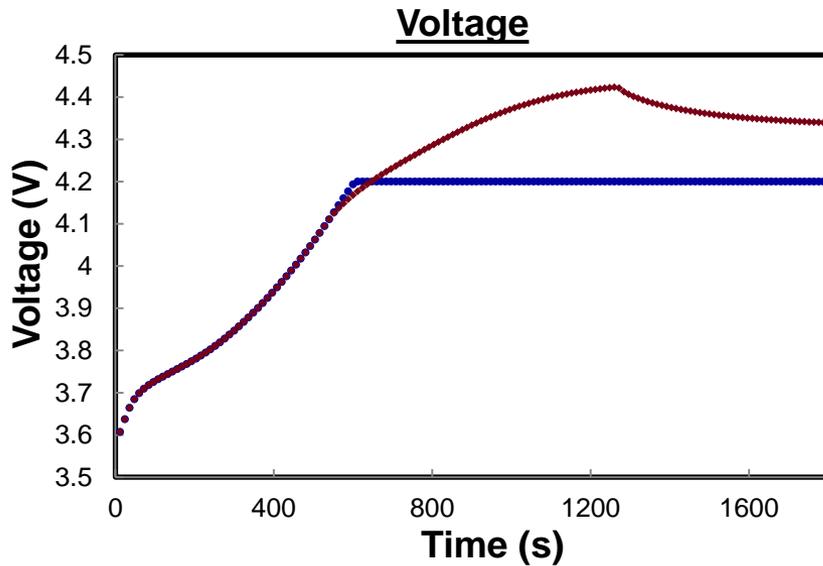
Reduces
into

Final System
30-50 DAEs

(P.W.C. Northrop+, JES, 2011)

(Patent US20140136169)

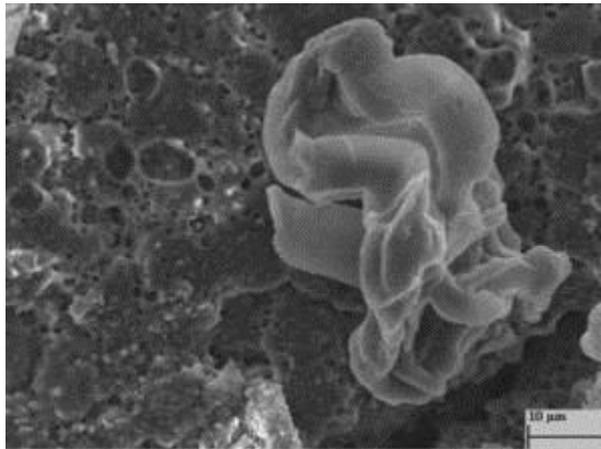
Which charging protocol is better?



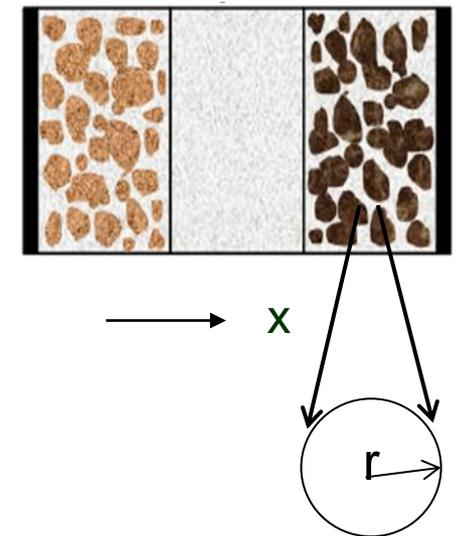
- **Protocol 1 reaches 4.2 V first and stays there**
- **Protocol 2 reaches 4.2 V at a later time but goes well above 4.2 V**

Lithium plating

- **Negative potential at the anode causes lithium to plate**
 - Anode overpotential < 0



Lithium deposited on the surface of a graphite electrode



(www.ifw-dresden.de/de)

Optimal control for charging – avoid plating

• Problem formulation

$$\max_{i_{app}(t)} Q = \int_0^{t_f} i_{app}(t) dt$$

Maximize charge stored

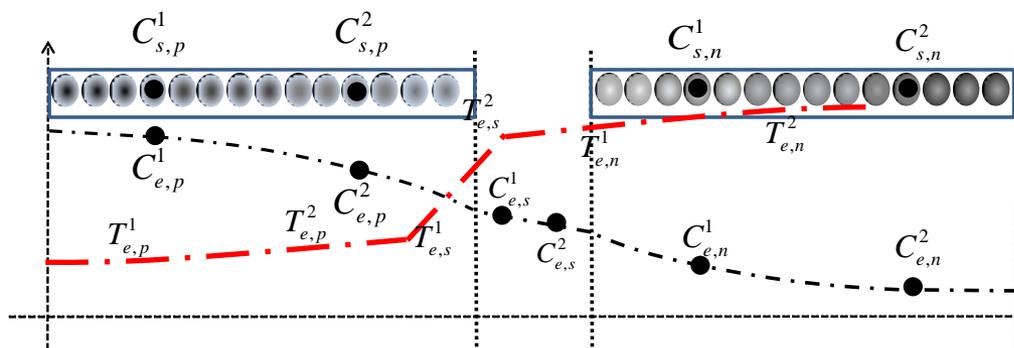
such that

$$0 \leq i_{applied} \leq i_{max} (=3.5C);$$

$$\eta_{plating} > 0$$

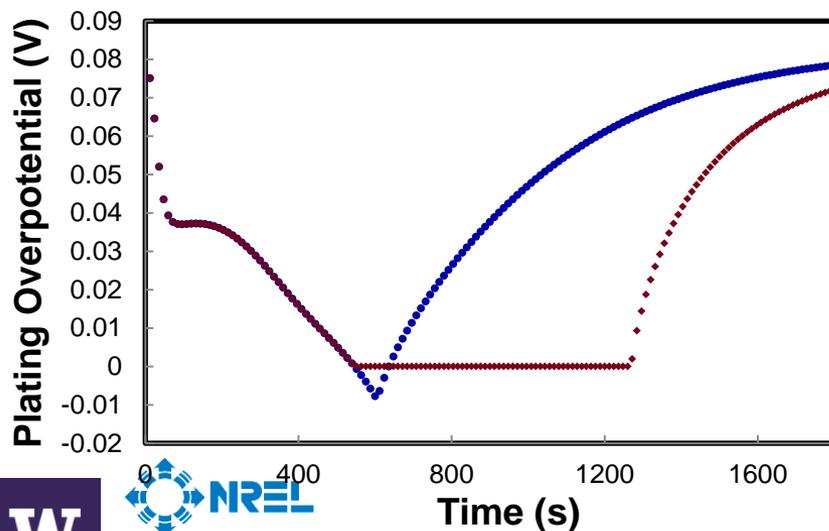
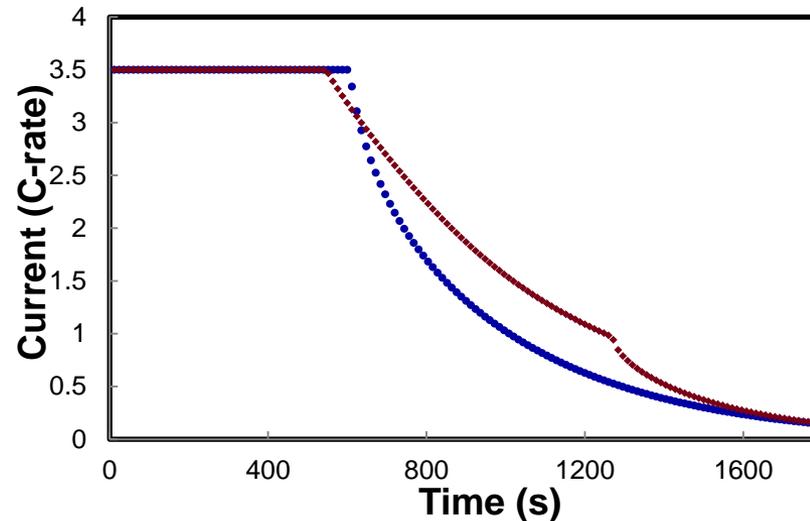
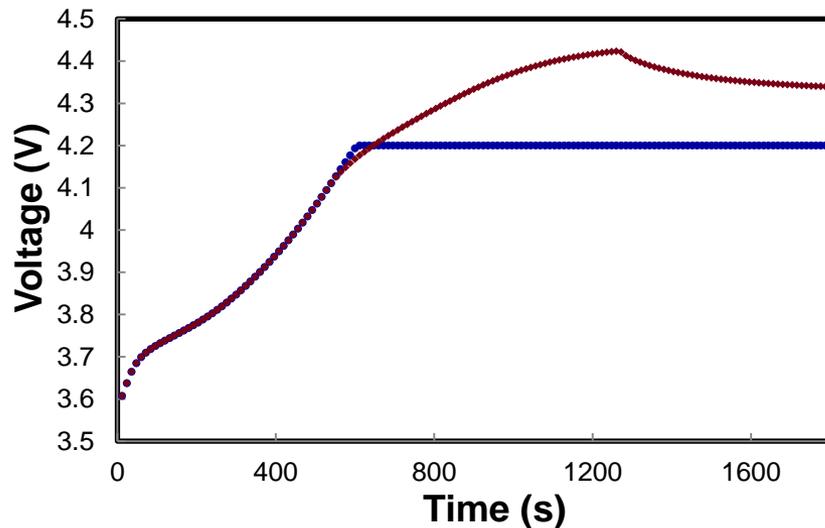
• Model description

– Reformulated model



(Patent US20140136169, “Systems and methods for improving battery performance;” P.W. C. Northrop+, JES, 2014)

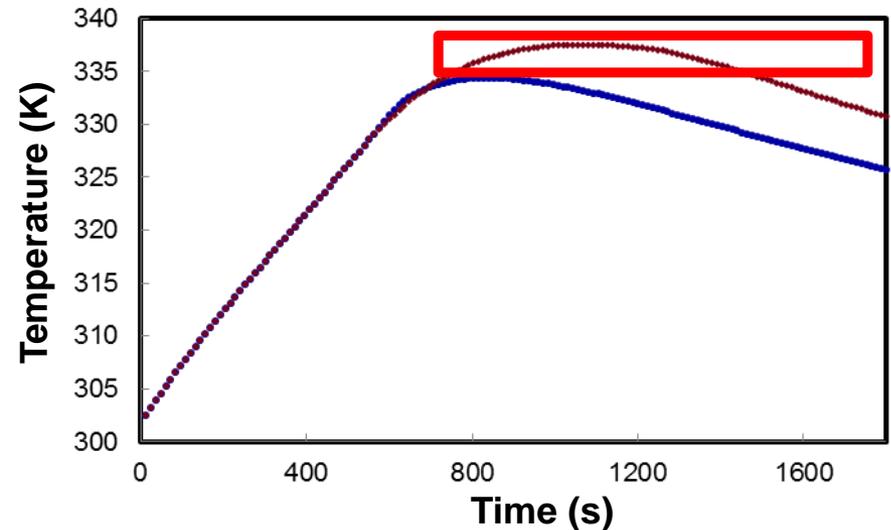
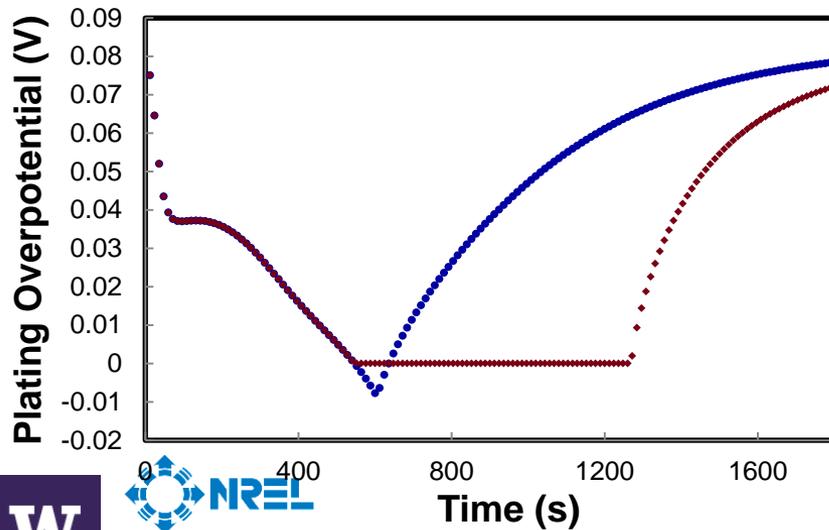
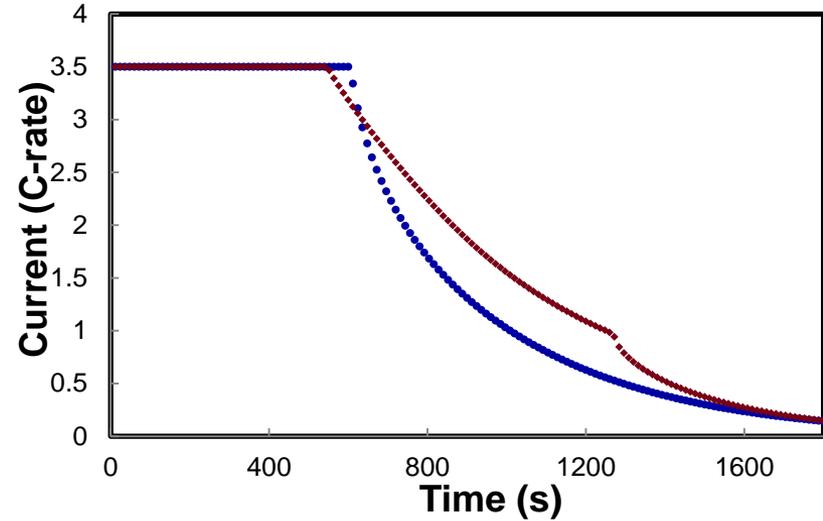
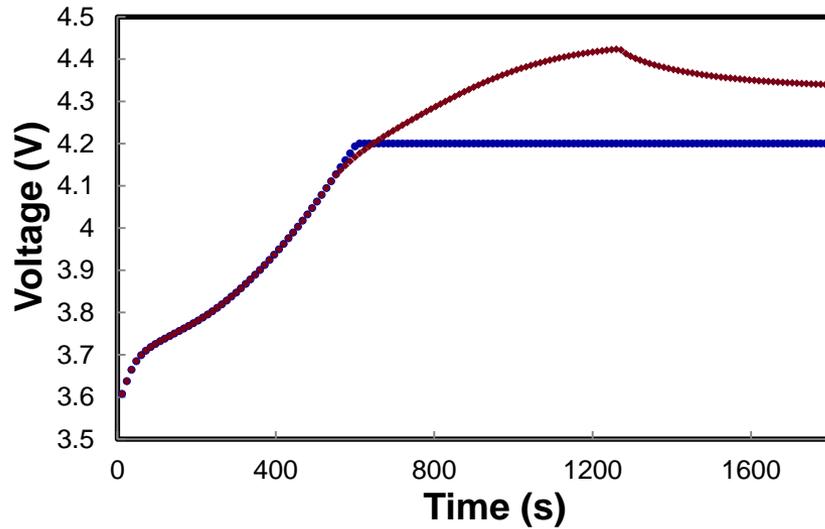
Which charging protocol is better?



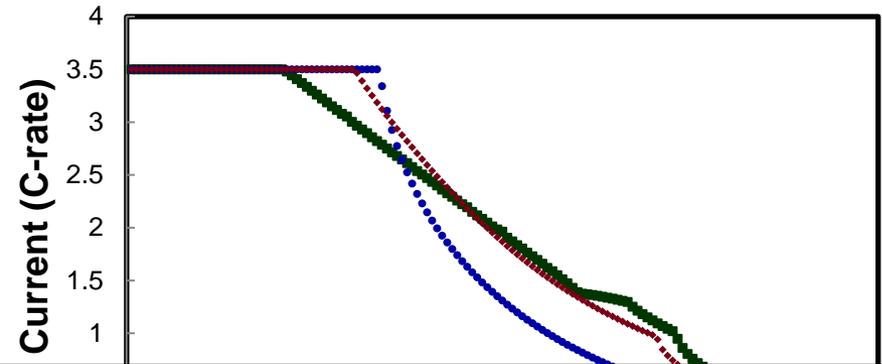
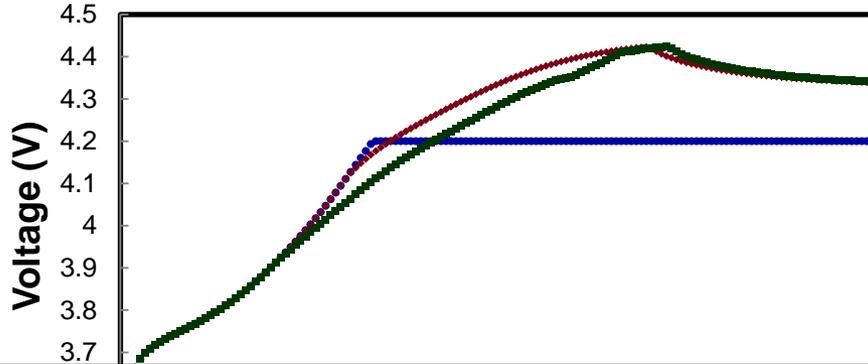
- Protocol 1 reaches 4.2 V first and stays there
- Protocol 2 reaches 4.2 V at a later time but goes well above 4.2 V
- Protocol 2 avoids lithium plating (optimal control)

P.W-C. Northrop +, JES (2014).

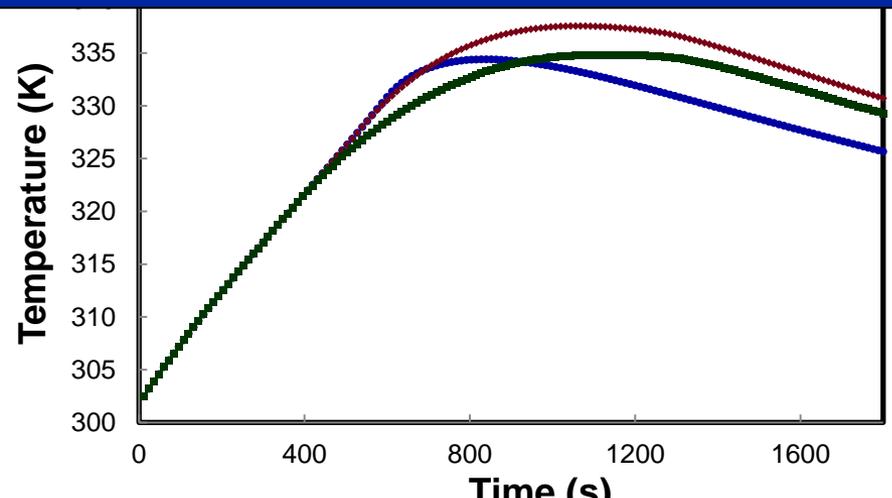
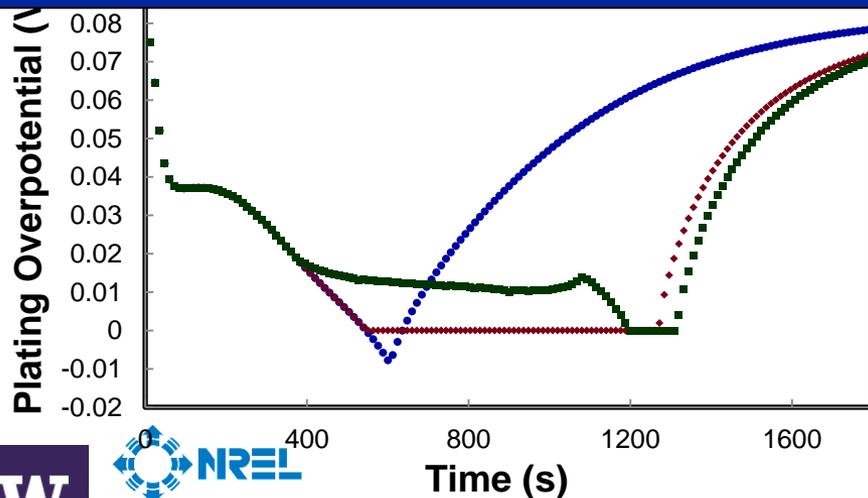
Is avoiding plating enough?



Determining optimal charge profile



Optimal fast charge yields 10.5% increase in SOC with no lithium plating or effects from temperature rise



User Interface

Battery Charging Optimizer



Optimal Battery Charging Protocol

A model-based optimization for maximizing State-of-Charge

M.A.P.L.E. Modeling, Analysis, & Process-control Laboratory for Electrochemical systems

Choose inputs for each system parameter below:

<u>Temperature</u>	<u>Max Current</u>	<u>Charge Time</u>	<u>Min Cycle Life</u>
310 K	1.0 C	30 min	500
<input type="radio"/> 310 K	<input type="radio"/> 1.0 C	<input type="radio"/> 30 minutes	<input type="radio"/> 500 cycles
<input type="radio"/> 315 K	<input type="radio"/> 1.5 C	<input type="radio"/> 45 minutes	<input type="radio"/> 750 cycles
<input type="radio"/> 320 K	<input type="radio"/> 2.0 C	<input type="radio"/> 1 hour	<input type="radio"/> 1000 cycles
<input type="radio"/> No bound			

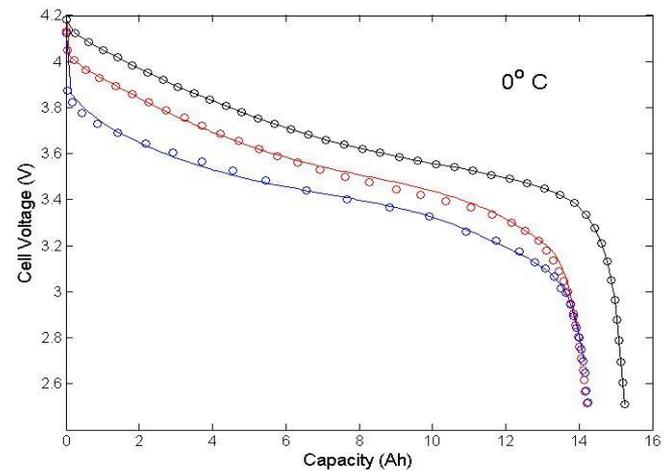
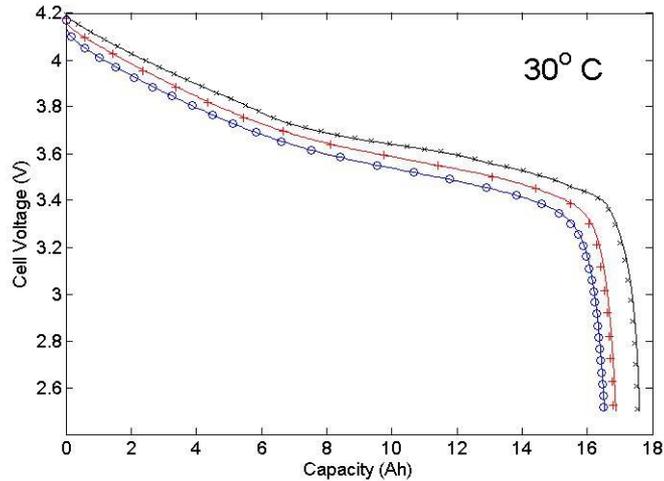
Run Charge Optimization **Help and Instructions**

PI: Venkat Subramanian vsubram@uw.edu

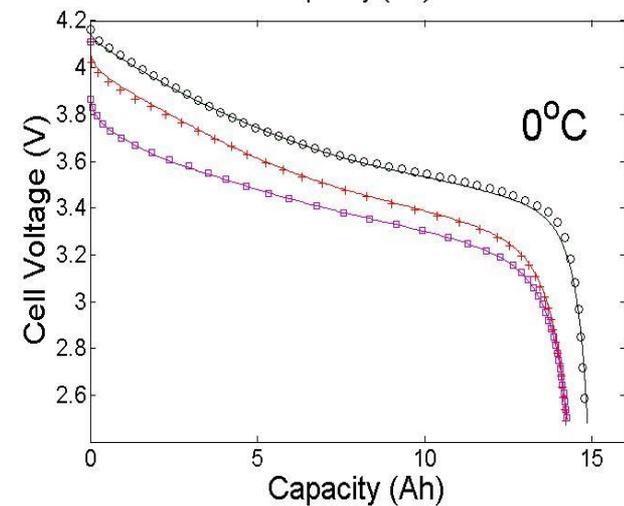
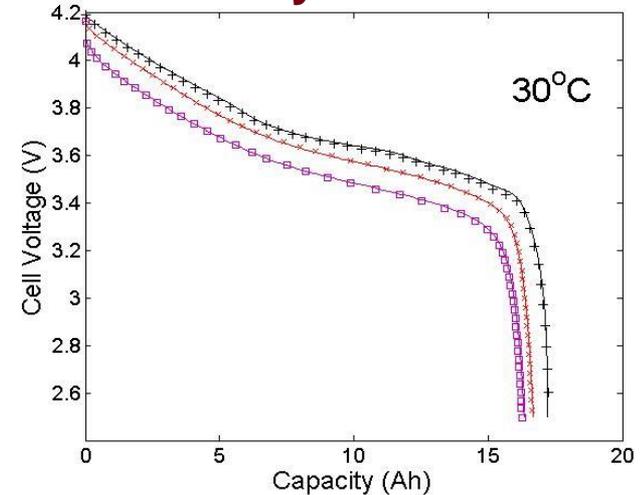


Model Vs Data (Case 0)

Fresh Cells

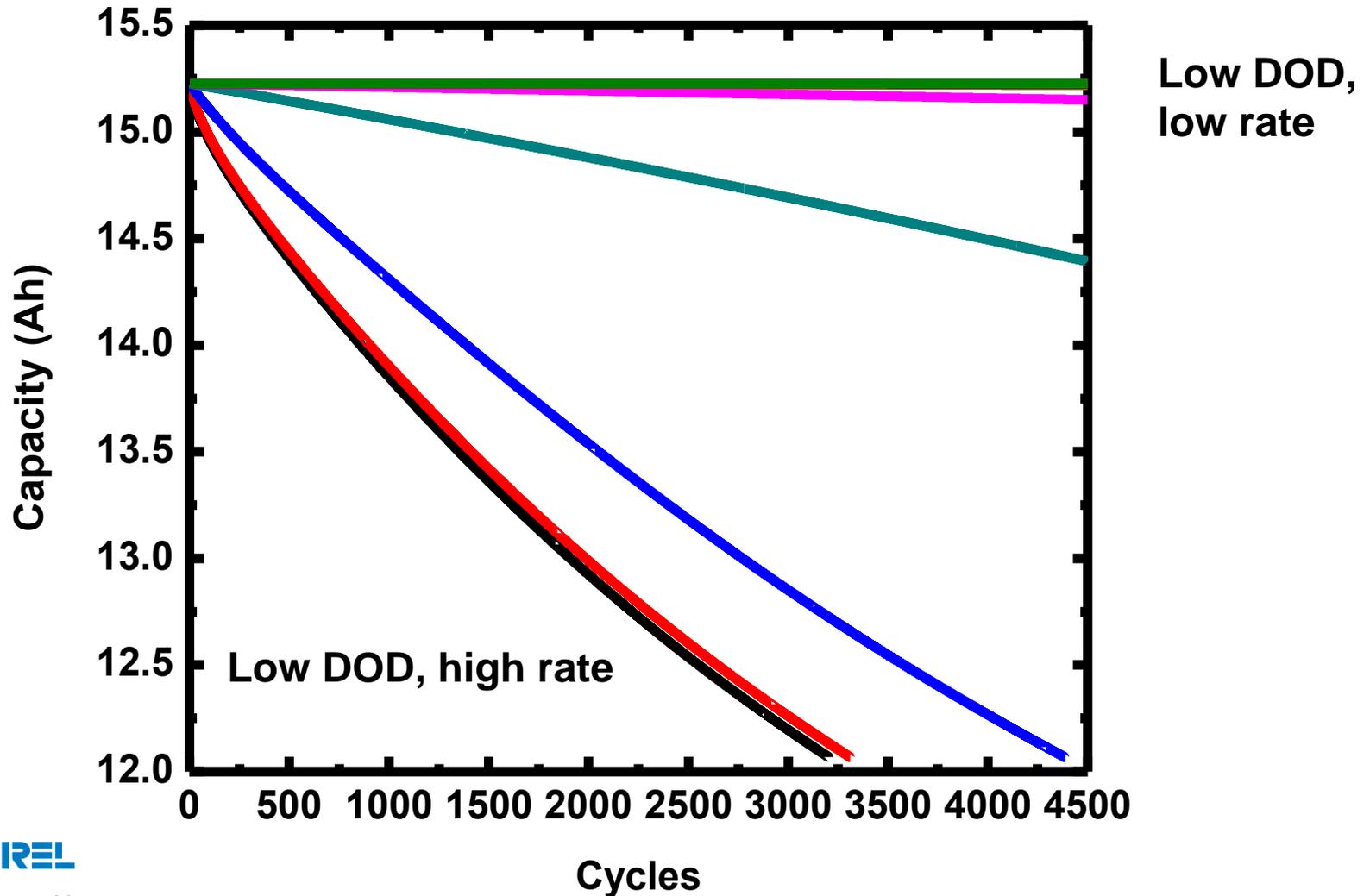


Cycled Cells

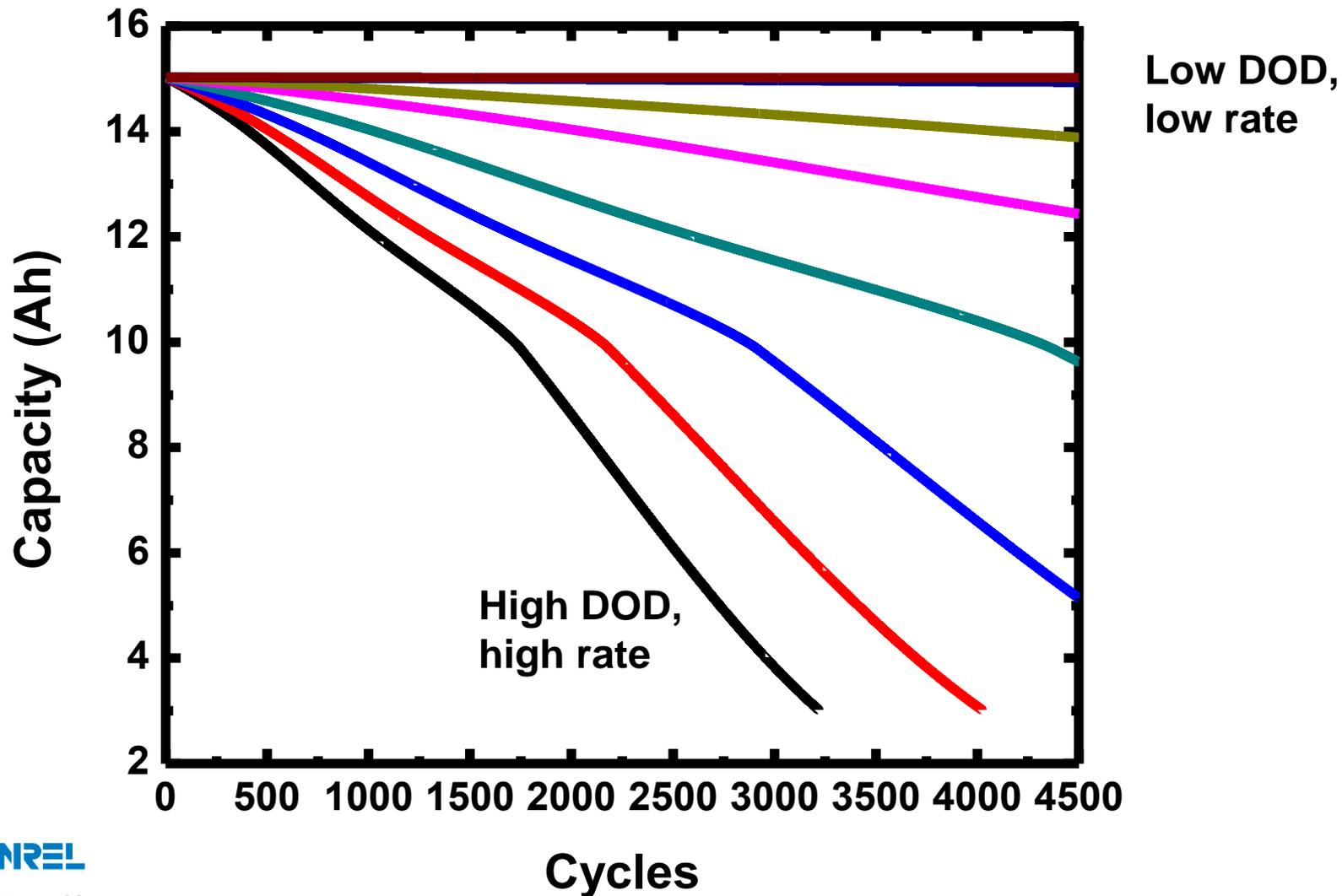


- Capacity check between 4.2 and 2.5 V
- 0.1C to 3C rate discharge

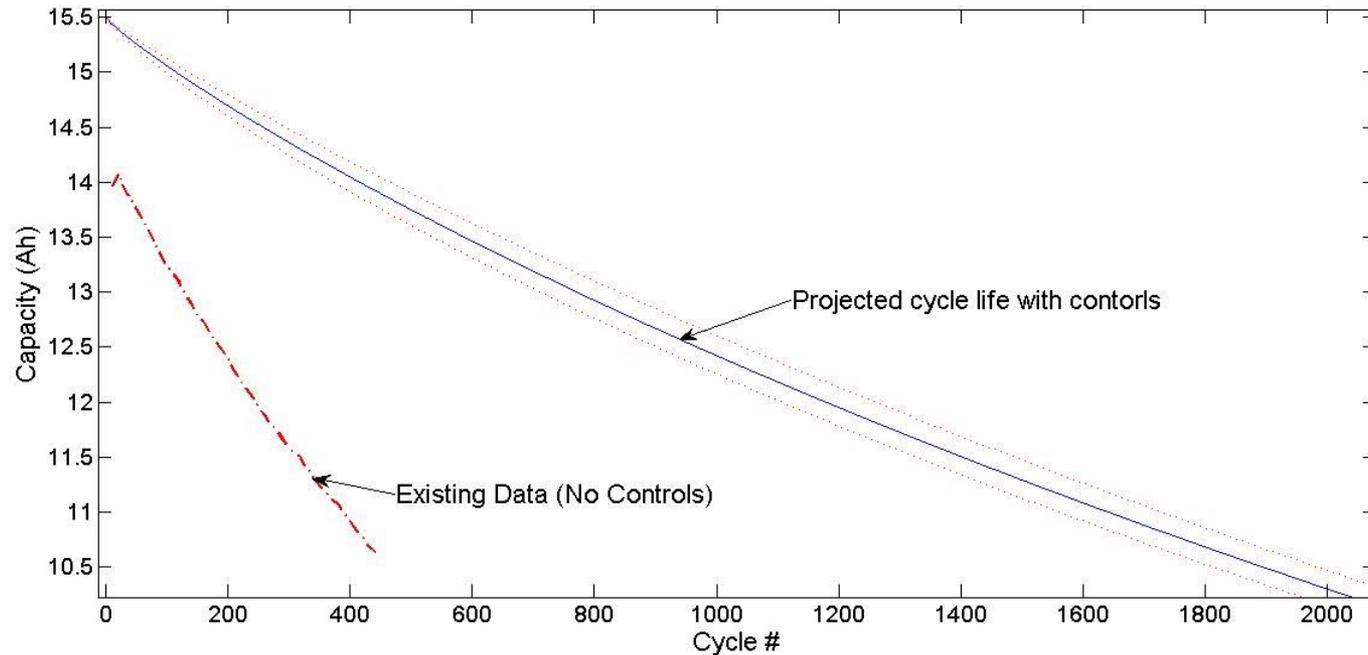
Life prediction (Best case, room temp)



Life prediction (all scenarios)

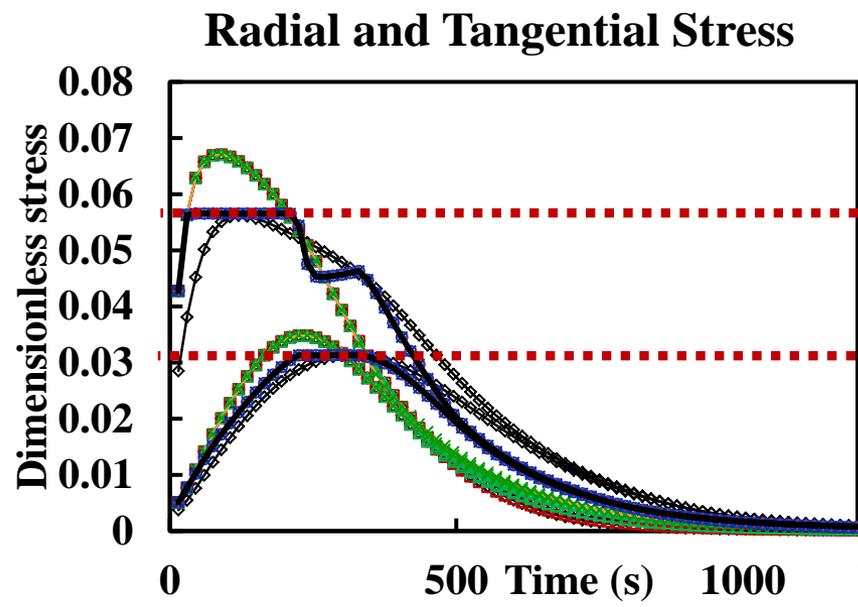
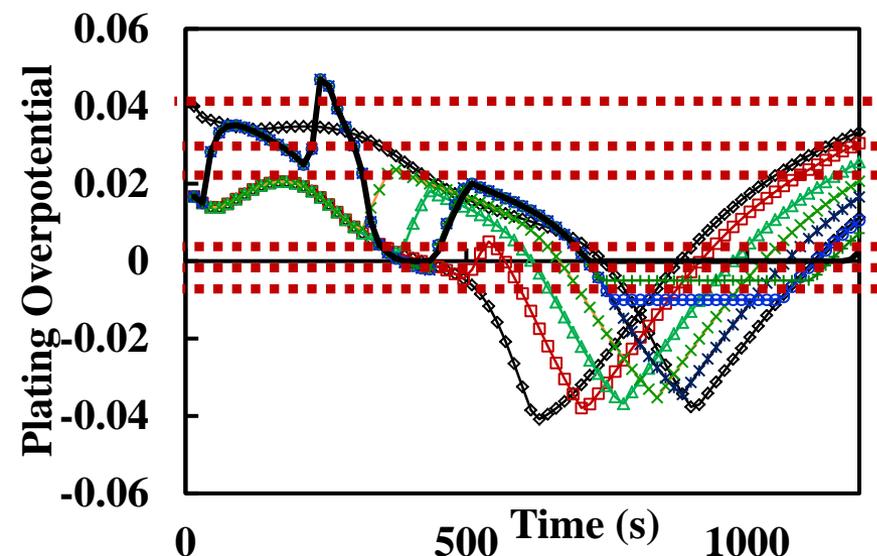
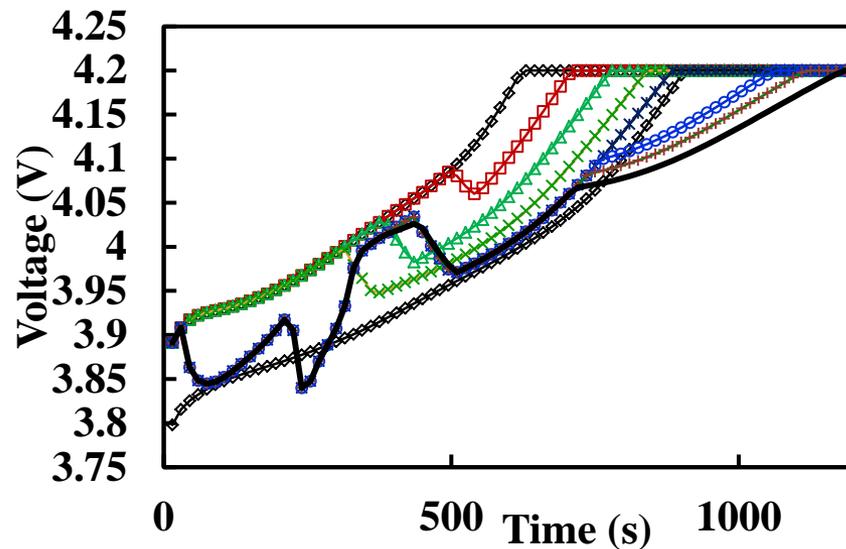
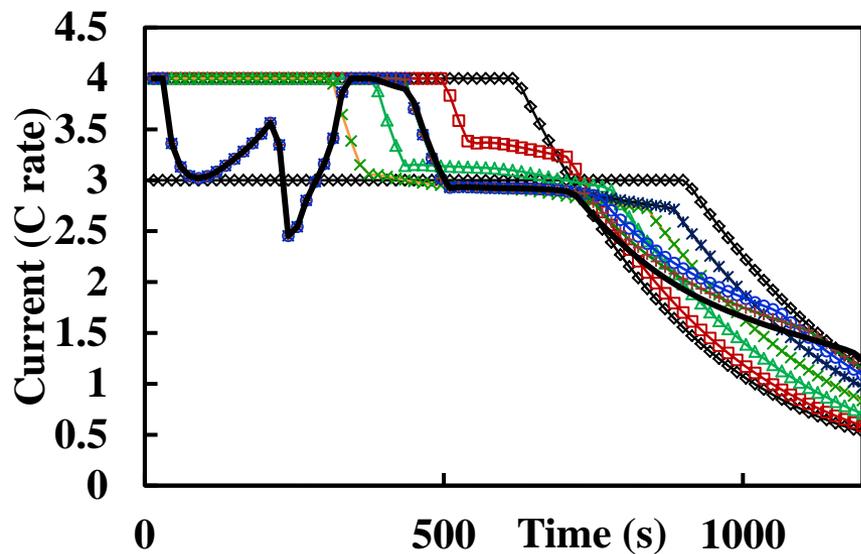


Extending Operational Window



- During 2C charge/5C discharge at 0 deg. C, the cell capacity drops to less than target usable capacity within 500 cycles.
- MPC will enable longer cycle life.

Optimal charging profiles – for newer materials



Publications and Codes

- **Lawder, M., et al. “Model-Based SEI Layer Growth and Capacity Fade Analysis for EV and PHEV Batteries and Drive Cycles” *J. Electrochem. Soc.*, 161(14), A2099-A2108 (2014)**
- **Northrop, P.W.C., et al. “Efficient Simulation and Reformulation of Lithium-Ion Battery Models for enabling Electric Transportation” *J. Electrochem. Soc.*, 161(8), E3149-E3157 (2013)**
- **Suthar, B., et al. “Optimal Charging Profiles with Minimal Intercalation-Induced Stresses for Lithium-Ion Batteries Using Reformulated Pseudo 2-Dimensional Models” *J. Electrochem. Soc.*, 161(11), F3144-F3155 (2014)**
- **Code for *optimally* charging batteries and GUI for battery optimization are available at www.depts.washington.edu/maple**