



Understanding Battery Safety and Durability Issues from Physics-Informed Data-Driven Modeling

Dr. Jun Xu, Associate Professor

Department of Mechanical Engineering, University of Delaware



Email: junxu@udel.edu

Energy Mechanics and Sustainability Laboratory (EMSLab)



My neighbor Tonny asked: What not buy an electric car?

Range Anxiety?



Higher energy density e.g., Lucid Air: 600 mi Larger pack Larger e.g., Tesla: super fast charging charging, 80% in 40 mins current Limited progress.

Charging time?



Autos & Transportation

Safety!



Reuters







Energy Storage Materials, 2020

5/15/2024





FIRST MULTIPHYSICS MODELING





1D battery model Short circuit model Anode Cathode Separator Anode Cathode $U^{1D} = U^{St}$ Separator $\varepsilon^{l} \frac{\partial c^{l}}{\partial t} = -D_{eff}^{l} \nabla c^{l} + \frac{i^{l} t_{+}}{F}$ $A^{1D}i^{1D} = I^{St} \qquad \vec{E} = -\vec{\nabla}\phi \qquad \vec{j} = \kappa^{St}\vec{\nabla}\phi$ $Q_{St} = \frac{\vec{j} \cdot \vec{j}}{\kappa^{St}}$ $i^{l} = -\kappa_{eff}^{l} \nabla \phi^{l} + \frac{2\kappa_{eff}^{l} R_{g}T}{F} \left(1 + \frac{d\ln F}{d\ln c^{l}}\right) \left(1 - t_{+}\right) \nabla \ln c^{l}$ $T_a^{Tm}(t) = T_a^{1D}(t)$ Q_{St} $x_a(t) = x(t)$. $c(\sigma_1(t) - \sigma_3(t))$ Thermal model $T^{Tm}(t) = T^{Tr}(t) \qquad T^{Me}(t) = T^{Tm}(t) \qquad V^{Me}(t) = v^{Tm}(t)$

Thermal runaway model

 Q_{1D}

 Q_{tr}

$$Q^{sei} = H^{sei}W_c A^{sei} \exp\left[\frac{-E_a^{sei}}{R_g T}\right] c_{sei}$$
$$\frac{dc_{sei}}{dt} = -A^{sei} \exp\left[\frac{-E_a^{sei}}{R_g T}\right] c_{sei}$$

Mechanical model $\rho \frac{\partial^2 \mathbf{u}}{\partial t^2} = \nabla_X (F_L S) + F_V$ plastic Fixed Force

Electrochimica Acta, 2017

5/15/2024

Multiphysics modeling strategy





0.8

0.6

0.4

0.2

 $\times 10^{7}$

0.8

0.6

0.4

0.2

0







J. Mater. Chem. A, 2021

A BRIEF SUMMARY







• Triggering of ISC \rightarrow • Evolution of ISC \rightarrow • Triggering of TR



PHYSICS-INFORMED DATA-DRIVEN MODELING









HOW CAN WE TELL?



Data-driven classification



2024/5/15

CLASSIFICATION RESULTS



Data driven safety risk prediction



- 293,990, 46,600, and 180,768 samples were generated and used in the training of C0, C1, and C2,
- □ 10 % of the samples will be used as test samples, 5 randomly splitting.



Confusion matrix



J. Power Sources, 2022

SHORT-CIRCUIT RESISTENCE





J. Energy Chemistry, 2023

SHORT-CIRCUIT RESISTENCE





Vehicle Energy & Safety Laboratory (VESL)



- Battery safety and durability is a highly complicated problem
- Data-driven has demonstrated the strong capability to solve complex system problem
 - Predict safety risk
 - Classify safety status
 - Predict the short circuit resistance (determinant factor)

What is next?





GRAND OPPORTUNITIES

- Establish testing methodology for battery safety and durability
 - \circ Work with FM Global
- Establish shared database
 - Work with some universities and companies
- Room for improvement
 - Develop fast, safe, low-cost testing/characterization
 - $\circ~$ Dive deep into fundamental mechanism and develop new physics-based models
 - $\circ~$ Deal with multiscale descriptions
 - \circ For ML
 - $\circ~$ How to deal with insufficient data or missing data
 - $\circ~$ How to deeply interact data-driven and physics
 - \circ $\,$ How to use ML in testing and characterization $\,$
 - How to use LLM in collecting data and feed them into the database?
 - 0 ...



THANKS



Special Thanks: Dr. Yikai Jia (former EMSLab PhD/postdoc) Dr. Wenquan Lu (ANL) Dr. Donal Finegan (NREL)

Contact: **Prof. Jun Xu** Email: junxu@udel.edu