Recent Advances in the Intercalation Battery Mechanistic Modeling Approach

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- Introduction While relatively young, the mechanistic modeling approach has proven **extremely versatile and effective** for Lithium-ion battery diagnosis and prognosis. It is a **backward-looking modeling approach** relying on **digital twins** built from **different matchings of the electrode individual voltage responses**. The digital twins enable quantification of the degradation modes and open the gate for material-based diagnosis and prognosis without the need for complex models. The method allowed to conceptually explain counter intuitive concepts such as hidden degradation mechanisms, overdischarge induced capacity, and the sudden apparition of knees. Looking forward, the approach can still be improved and some **new directions** will be detailed here for the first time with **proof-of-concept simulations**. This includes more varied synthetic datasets, **blended and inhomogeneous electrodes** and packs, voltage fade, as well as **calculations outside of constant current**. These new features could allow a much wider field implementation of better diagnosis and prognosis tools for deployed systems. Cell degradation emulation using Digital Twins Degradation mechanisms vs. degradation modes LAMs are emulated by changing the size of one electrode (or component of the blend) compared to the other(s) and by adjusting the rate to account for changes in current density. Degradation modes refer to the impact of a mechanism rather than its root cause. Every degradation mechanism will impact the amount of material able to react, the amount of lithium able to go back and forth between the electrodes, and/or the mponent of a blend) at different rates. t only requires **few parameters**: The **loading ratio (LR)** that corresponds to the electrodes capacity ratio LLI is emulated by increasing the offset. This is because LLI is increasing the slippage between the electrodes. As such, three main families of degradation modes can be defined: loss of active material (LAM), loss of lithium inventory (LLI), and kinetics changes. LAMs and kinetic limitations must be decomposed further at the electrode or electrode component levels for blends. The **offset (OFS)** that corresponds to the electrode slippage, The **rate** for each electrode (or component of the blend), **Ohmic resistance** changes are emulated by translating the voltage responses up or down. e resistance for each electrode (or component of the blend) Dyebert
 Directory Faradic resistance changes are emulated by adjusting the rate and the resistance together. To simulate faster/slower kinetics data at a slower/faster rate is used then polarization is adjusted so that the overall resistance stayed the same. Blends are currently emulated by summing the pondered IC ****** Kinetic accommodation principles Example of a NMC+LMO blend LiMn_Ni_Co_O The second 0 A B B B B The state 3.6 Li plating & hidden mechanisms **Blends & Voltage fade** With a mechanistic view, plating can only originate from two pathways: NE full before PE is empty or NE potential < L¹/L¹ potential Plating does not induce capacity loss but turns the NE into a blended electrode. Irreversible plating = LLI and different reversibilities can be simulated. The mechanistic approach was proven efficient to simulate and forecast plating as commercial cells started to use blended electrodes, her on the positive or negative electrodes. approach in the present model implies that the current has 600 to be constant on both component of the blend throughout the 400 LMO 400 NCA Recent work by Heubner *et al.* showed that the **reality is more** complex and that three case figures needs to be considered: separated voltage responses (e.g. LFP+LMO) 3.8 PE mostly overlapping responses (e.g. NMC +NCA) partially overlapping responses (e.g. NMC+LMO or Gr+SiOx) NCA LMO 20% LAM OCV after aging
 C/10 after aging
 C/10 initial 20% LAM_{NE} sent approach is **not accurate for 2nd and 3rd cases** 0.06 50% reversibility 100% reversibility mechanistic approach can be upgraded with the ementation of a **paralleling model** to be more accurate. 0.04 60 60 80 80 0% LAM., 0% LAM. 20 40 60 80 Resistance x10 -10% rate capability Voltage (V) 3.5 A.; Liaw, B.Y. J. Power Sources 2016, 321, 36-46, doi:10.1016/j.jp Blend simulations also allow to remove one of the approach solution 2! Non-constant current 20 40 60 80 100 20 80 If variable, phase transitions or solid solutions evolutions could be taken into consideration as blends with varying ratio. 40 60 SOC (%) While most usage of the mechanistic approach is done for constant current, this is not a prerequisite. Other duty cycles could allow to investigate impact of degradatior SOC (%) modes on **constant power or constant voltage steps**. It could allow to apply th approach on deployed systems with semi constant cycles or to optimize charge Inhomogeneous electrodes / packs Vith the manufacturing of larger and larger single cells and packs, the lectrochemical behavior of electrodes, or single cells, was shown to be chomogeneous. Paralleling several mechanistic models could allow to nvestigate the impact of cell-to-cell variations, inhomogeneities in legradation, localized defects, or lithium reservoirs on the response of Using the ability to simulate different rates, the mechanistic CCCV the full electrode using a **segmented cell digital twin** where each segment is individually parameterizable. simulate different duty cycles. By aggregating charge and discharge curves at different rates, data following a specific Constant Powei Appedito 1150 can be selected for each SOC and the full V vs. SOC response 00 500 (% can be constructed point by PV response 🛔 0 2 4 6 8 10 0 2 4 6 8 10 Time (1) Time (1) . 3.5 3.6 3.7 3.8 3.9 4 20 Acknowledgments **Conclusions & Perspective** Mechanistic Modeling Approach Validated Under validation The authors are grateful for Office Electrochemical responses Na-ion and other chemistrie: Advanced blends Voltage fade Inhomogeneous electrodes of Naval Research (ONR) funding L LAMS
- battery diagnosis and prognosis. Enables quantification of degradation modes with material-based prognosis without the need for complex models, protocols, nor instrumentation. Allows to explain counter intuitive concepts such as hidden degradation

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duty cycles

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