

Advanced Battery Management System Methods for Stationary Storage Applications

Preliminary results from the ISTORMY project

Active Battery Diagnostics (ABD)

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Battery State-of-Charge (SoC) & State-of-Health (SoH)

A coupled State-of-Charge & State-of-Health estimation algorithm based on a Dual Kalman Filter approach is proposed.

Hysteresis in LFP Cell ^c

First, performance tests are

Cea

The ABD system actively diagnoses battery health of real world applications through manipulation of bi-directional power electronics. The output of the ABD system feeds up-to-date parameters to other BMS

functions like the SoC, SoH and RUL

Ideal system to **harvest** high quantity + quality **battery degradation data!**



realized in the laboratory on the LFP and NMC cells that compose the ISTORMY battery pack in order to build n-RC equivalent two electric circuit models. A particular focus is made on the characterization of the hysteresis phenomenon of the open circuit voltage.



The "Interoperable, modular and smart hybrid energy storage system for stationary applications" (ISTORMY) project is funded by the Horizon 2020 program of the European Union. It presents an innovative Hybrid Battery Energy Storage System (HBESS) for grid integration, based on modular conception of the battery pack and power electronics. The goal is to optimize the system to increase its lifetime and reduce the Total Cost of Ownership (TCO). To that end, the Battery Management System (BMS) offers advanced diagnosis functionalities.

The ABD system measures battery health by applying a specific combination of current waveforms. These waveforms are designed to make **battery health parameters** like capacity and resistance more accurately measurable. A trade-off is struck between accuracy, precision and measurement time.



The ABD system provides high quantities of relevant battery health data for training of the data-driven battery RUL algorithm (see below).



Ageing tests are also performed on the cells to build a semi-empirical model of capacity loss that considers both calendar and cycling degradation mechanisms.

oltage



State-of-Health The estimator makes а continuous prediction of the capacity of the battery using the ageing model, and uses the output of the Active Battery Diagnostics as a periodic observation of the performance model parameters evolution.



It is important to note that the method can easily be adapted to any kind of Li ion cell chemistry (LFP and NMC in this work).





The iSTORMY system uses multiple different chemistries of battery cells, combining the strengths of each.

The iSTORMY hybrid energy storage system will be tested at the EDF microgrid demonstration location

Battery Remaining Useful Life (RUL)

A data-driven approach of battery Remaining Useful Life (RUL) estimation is presented, based on data harvested by the active battery diagnostics. The current preliminary method uses support vector regression in combination with a compressed historic record of the battery's usage (including stress factors).

When coupled with the State-of-Health algorithm, the error on State-of-Charge estimation remains stable along the battery's ageing. The electrical parameters of the two packs are tracked, providing **precise** and **robust** information to the Energy Management System.



Training data statistics	Laboratory data	Active Battery Diagnostic data
Number of battery cells in data	n ∼ 10-100	> 500
Cost of creating data	€€€	€
Usage pattern of data	Often CC or CCCV 0°C, 25°C or 45°C	Realistic real world use
RUL Pridiction b		
0.99 0.98 Usage pattern 1	 Measured Prediction 0.96 Set Prediction 0.95 Set Prediction 0.93 Set 	П the graphic to the left, th RUL algorithm has bee trained on data of usag pattern 1 and 3. but not c

The 2. usage pattern visibly algorithm struggles with the usage pattern it has not been trained on, yet is able correct when to retrained on new incoming data.

Additional training data harvested by the ABD system will likely significantly improve algorithm performance over time.

a Meulenbroeks, T. A. J. "Active Battery Diagnostics for Real World Applications", 2022 **b** Meulenbroeks, T. A. J. "Battery Remaining Useful Life Estimation Driven by Active Battery Diagnostics", 2022 c Werckle, C. "Improved Equivalent-Circuit model of Lithium-ion Iron Phosphate Cells", 2005 d Plett, G. "Dual and Joint EKF for Simultaneous SoC and SoH estimation", 2004

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