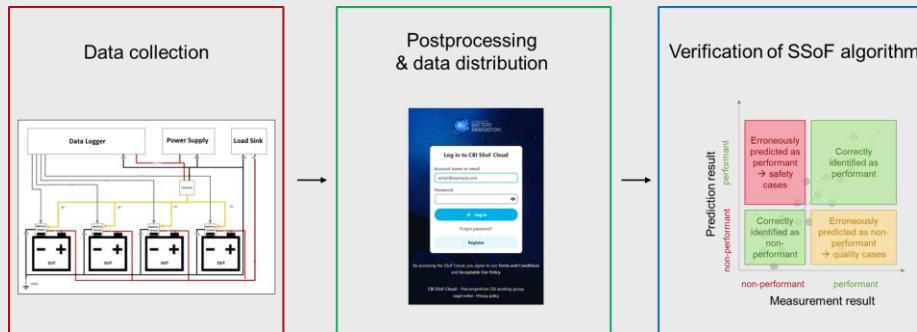


Statistical analysis of short-term prediction for Safety State-of-Function

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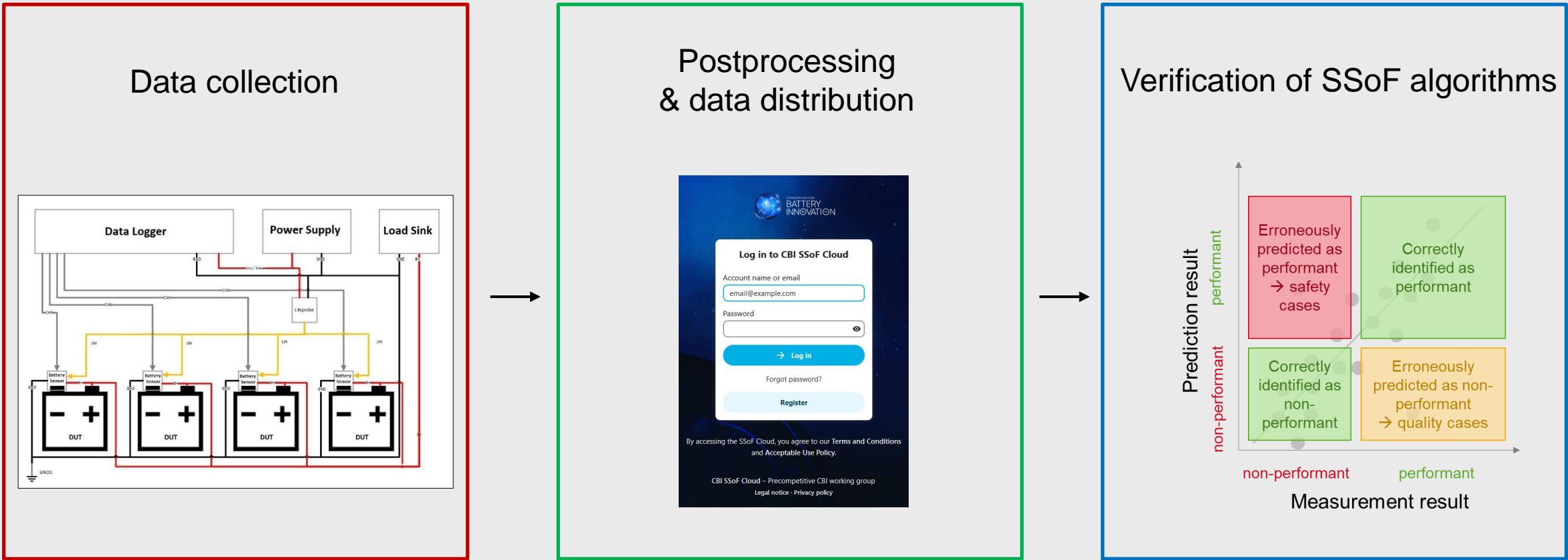
Jonathan Wirth
BatterielIngenieure GmbH

Eckhard Karden
Consultant, Consortium for Battery Information (CBI)



- CBI SSoF database
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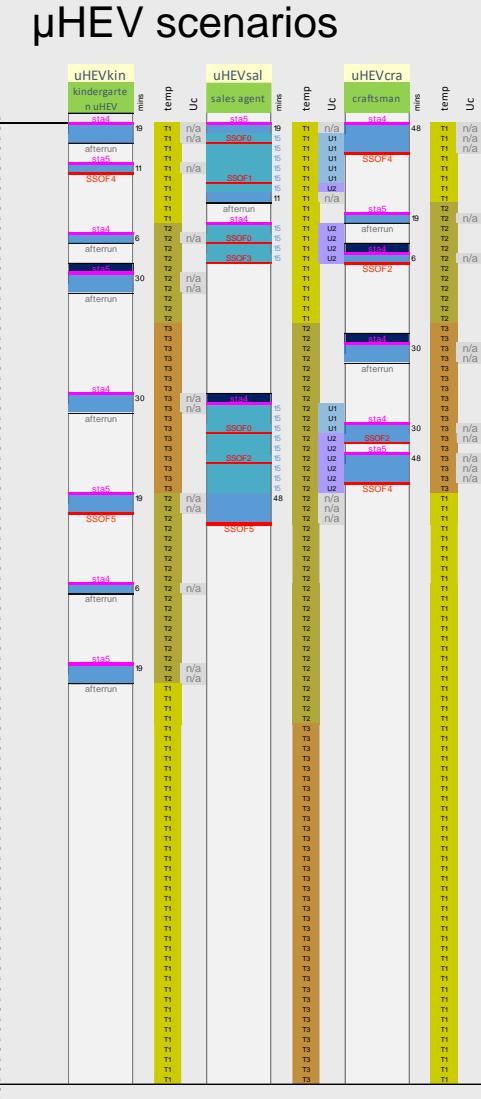
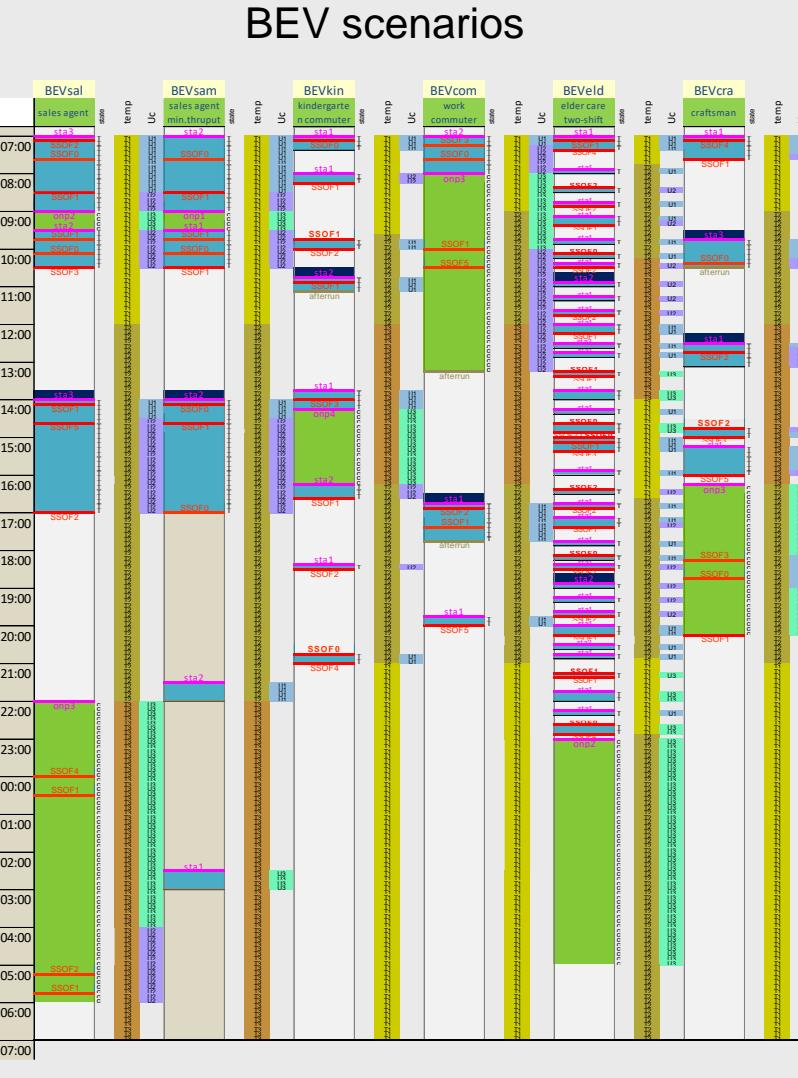
Steps for SSoF validation



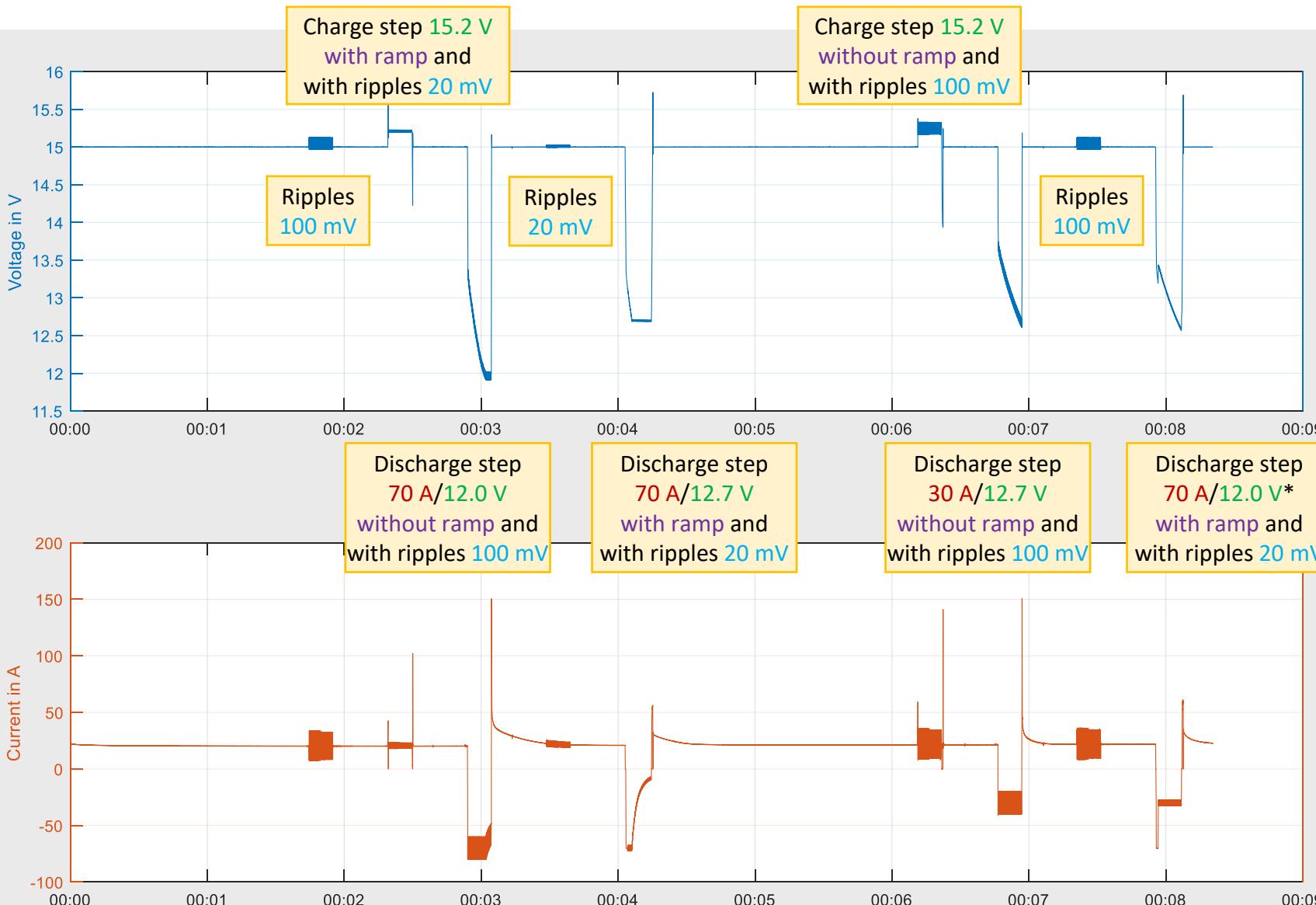
- 320 test days with > 3.000 SSoF, > 50.000 large signal stimuli, > 80.000 small signal stimuli as of now
- Dataset used in this study: 9 test days of one 70 Ah battery

Simulated real-world driving – cycle definitions

- **BEV and μHEV**
 - Charging cycle for BEV
 - Urban and extra-urban trips of different duration for μHEV (incl. stop/start)
- **Usage patterns:** Sales agent, kindergarten or work commuter, eldercare, craftsperson, ...
- **Climate conditions:** Temperature and generator/alternator setpoint adjustments for different scenarios



BEV charging cycle



Ripples, charge steps and discharge steps (with and without ramps)

Voltage setpoints:

- 15.2 V for charging steps
- 12.7 V and 12.0 V for discharging steps

Ripple amplitudes:

- 20 mV and 100 mV

Current limits:

- 150 A for charging steps
- 70 A and 30 A for discharging steps

* 30 A for ripples in this measurement

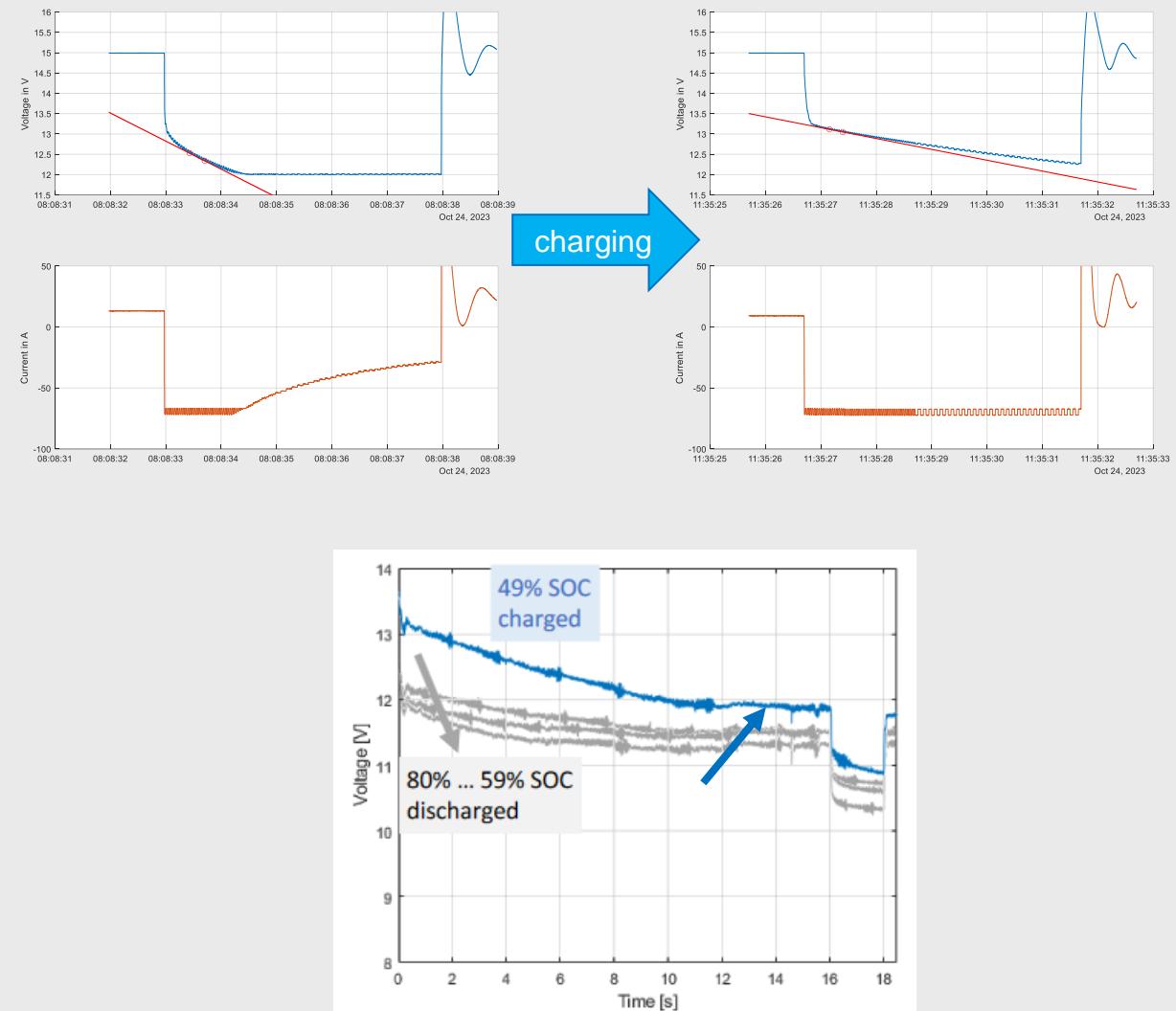
SSoF recovery out of charging condition

■ Effects accelerating transient recovery of SSoF over SoC recovery

- Pseudo-capacitance of positive electrode building up during charging
(cf. Eberhard Meissner, ALBA 2024)
- Discharge voltage stabilizes at elevated level compared to battery at equilibrium or freshly discharged
(cf. Grzegorz Pilatowicz, Dissertation 2017)

→ During charging, SSoF recovers very fast already over a few percent of SoC change

■ Passive equivalent circuit models without SoC parameterization will be used in this study

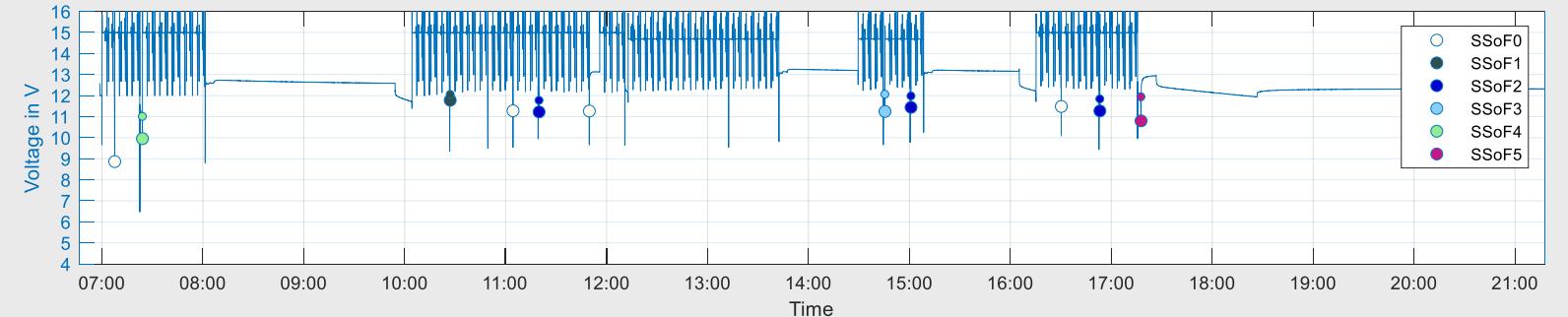


Outline

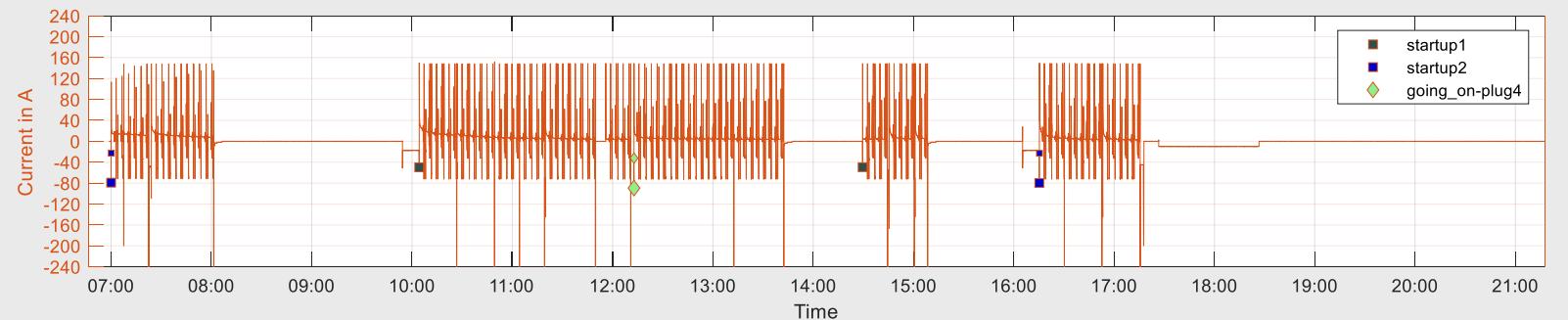
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Example of BEVsal

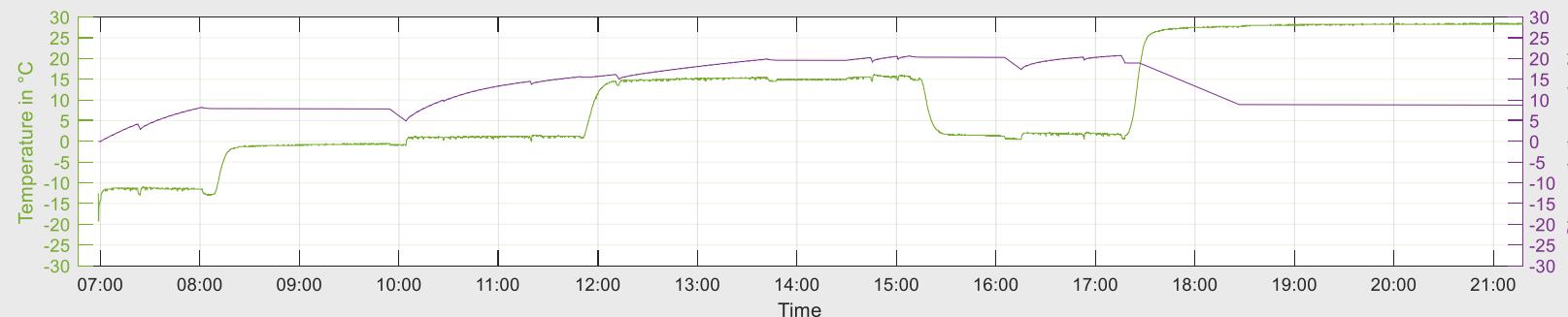
Combatec BEV Oct 23 (25) | 2023-10-24_07-56-53 | Measurement_1 | CAN1 | BEV_Day2



Battery voltage with
end voltages of SSoF

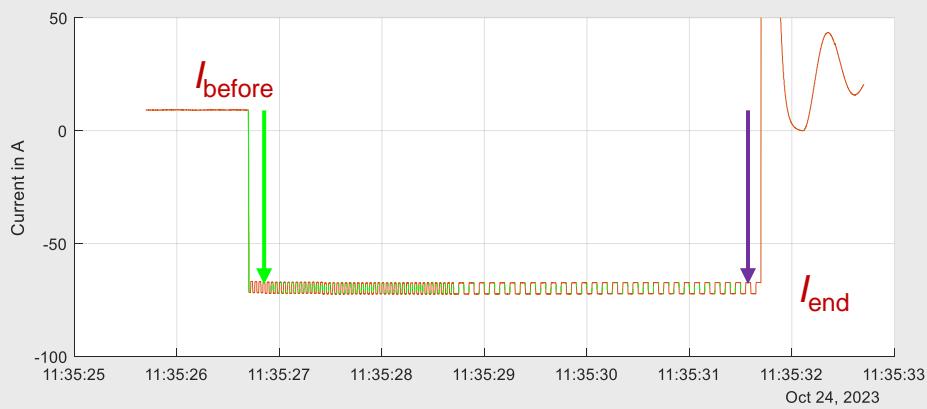
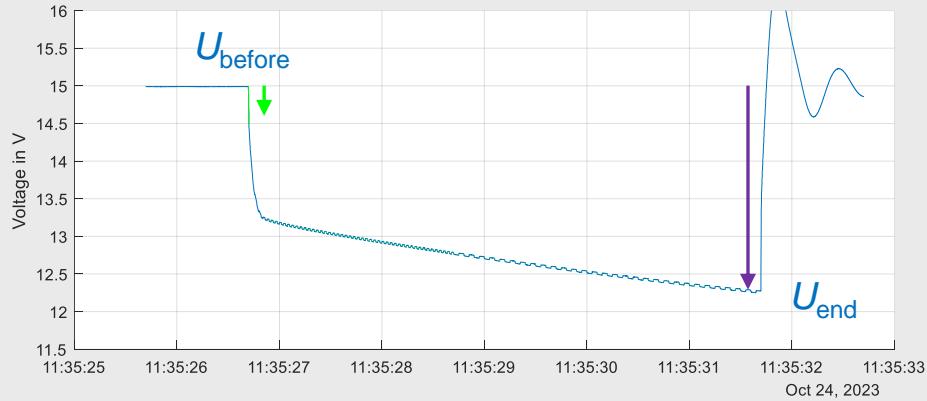


Battery current with
end currents of
startups, going on-plugs



Temperature and
charge balance
(fixed scales for comparison)

Model 1: Ohmic resistance model



- Specification dch_step1:
- 70 A for 5+ s*
 - 12.0 V voltage limit
 - superimposed ripples $\pm 10 \text{ A} / \pm 0.1 \text{ V}$

* duration depending on test bench control

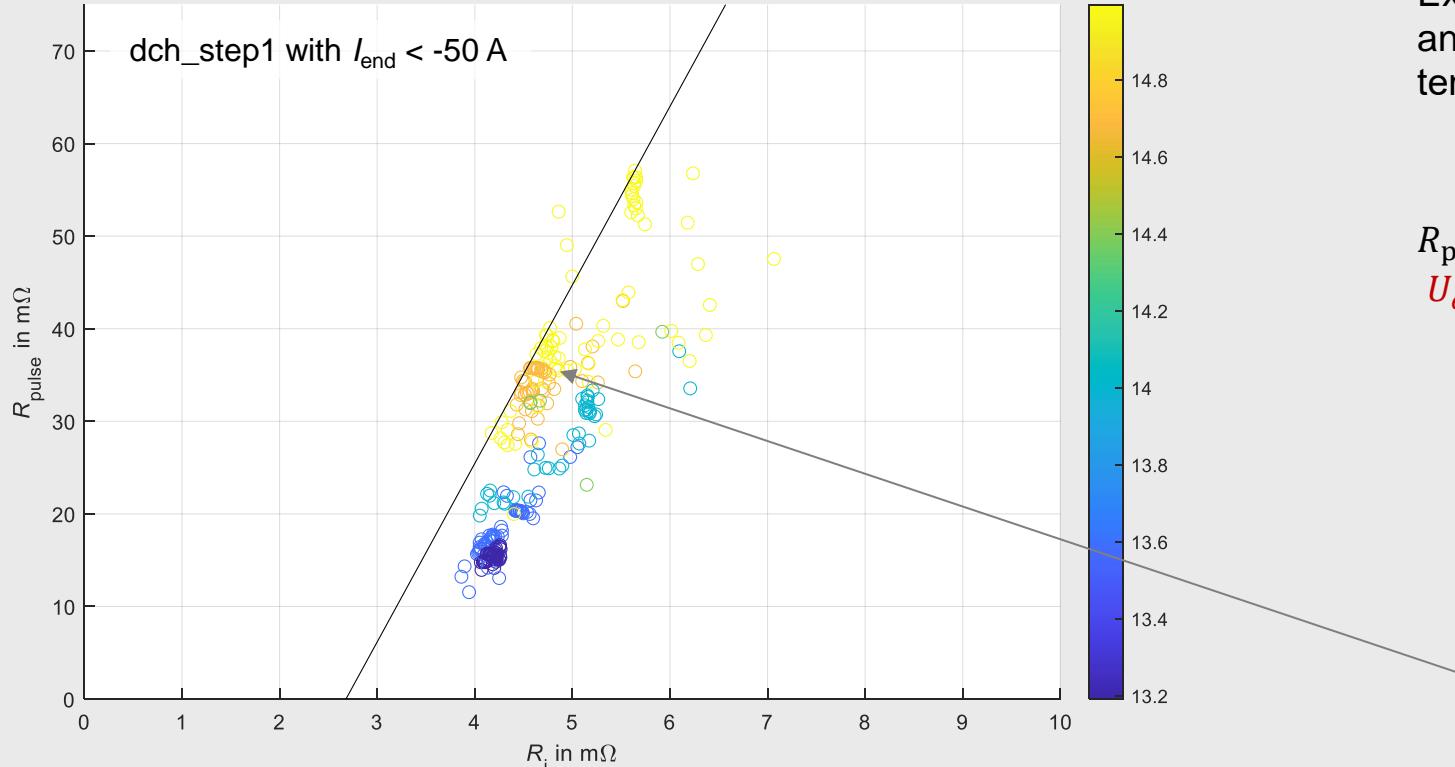
Ohmic drop based on ohmic resistance R_i from fast changes in ripples

Prediction of U_{end} needs further knowledge about ratio between pulse resistance R_{pulse} and R_i

$$R_{\text{pulse}} = \frac{U_{\text{end}} - U_{\text{before}}}{I_{\text{end}} - I_{\text{before}}} = f(U, I, T, \Delta t, \dots) \cdot \underbrace{(R_i - k)}_{= \text{const.}} \quad (\text{for this model})$$



Model 1: Parameter extraction

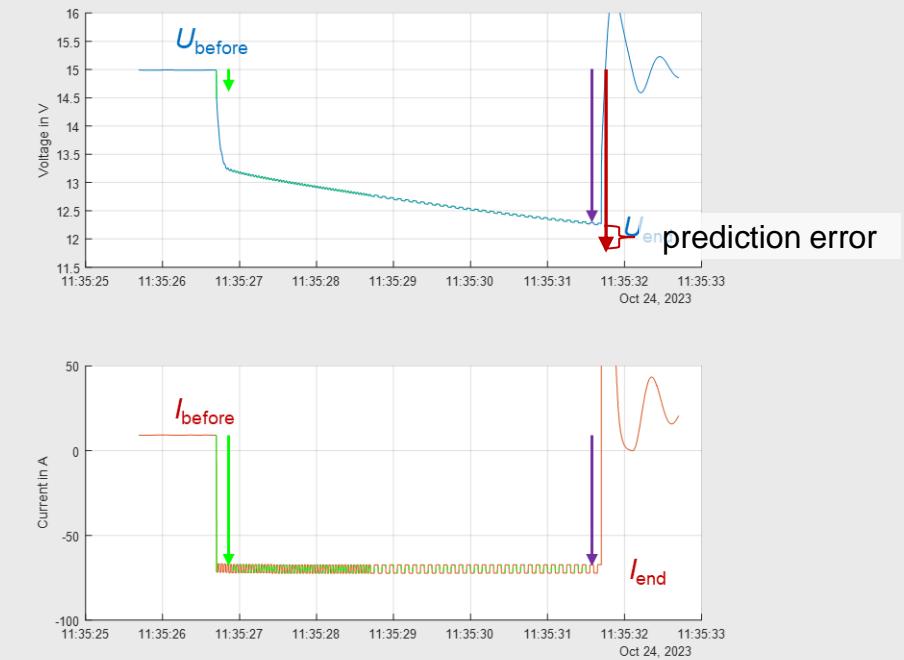


Example of ratio between pulse resistance R_{pulse} and R_i fitted for one battery, independent of temperature, charging voltage, SoC, ...

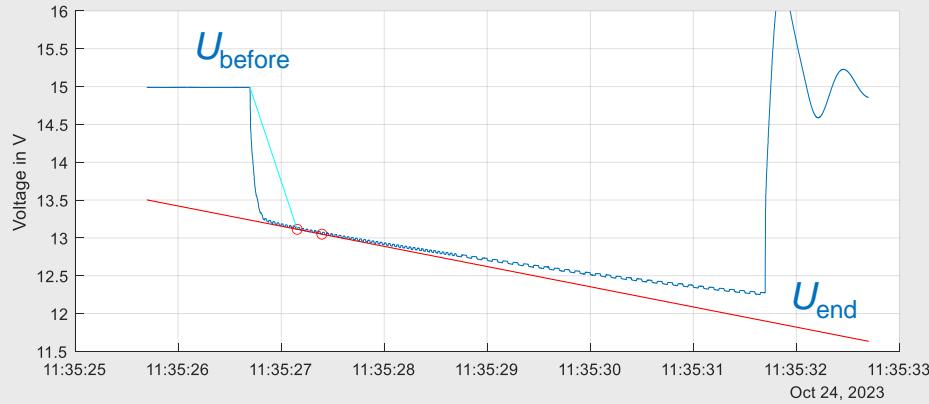
$$R_{\text{pulse,pred}} = (4.85 \text{ m}\Omega - 2.68 \text{ m}\Omega) \cdot 19.3 = 41.9 \text{ m}\Omega$$

$$U_{\text{end,pred}} = U_{\text{before}} + R_{\text{pulse}} \cdot (I_{\text{end}} - I_{\text{before}}) = 11.78 \text{ V}$$

=

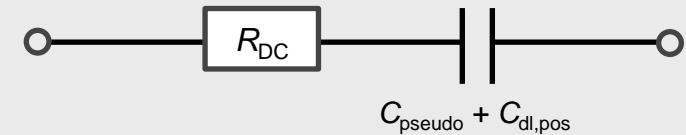
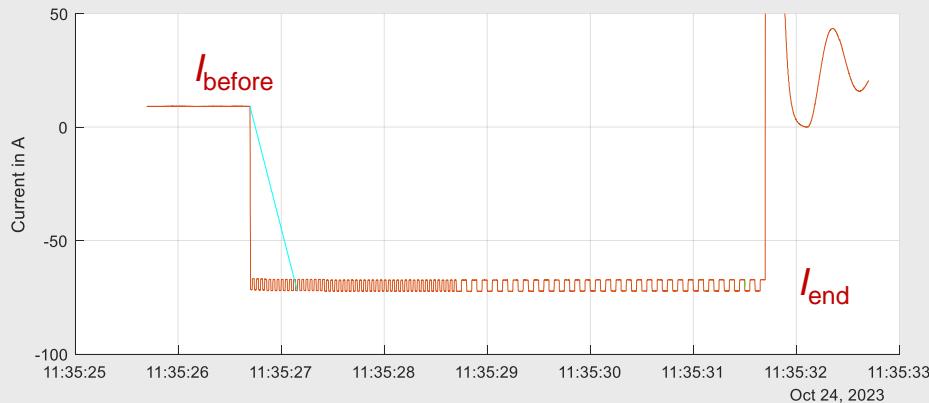


Model 2: Pseudo-capacitance model



R_{DC} after discharge of $0.0125\% \times C_n$
to include fast transients of neg. electrode

Slope $\Delta U / \Delta Q$ is dominated by pseudo-capacitive behavior of positive electrode
(cf. Eberhard Meissner, ALBA 2024)

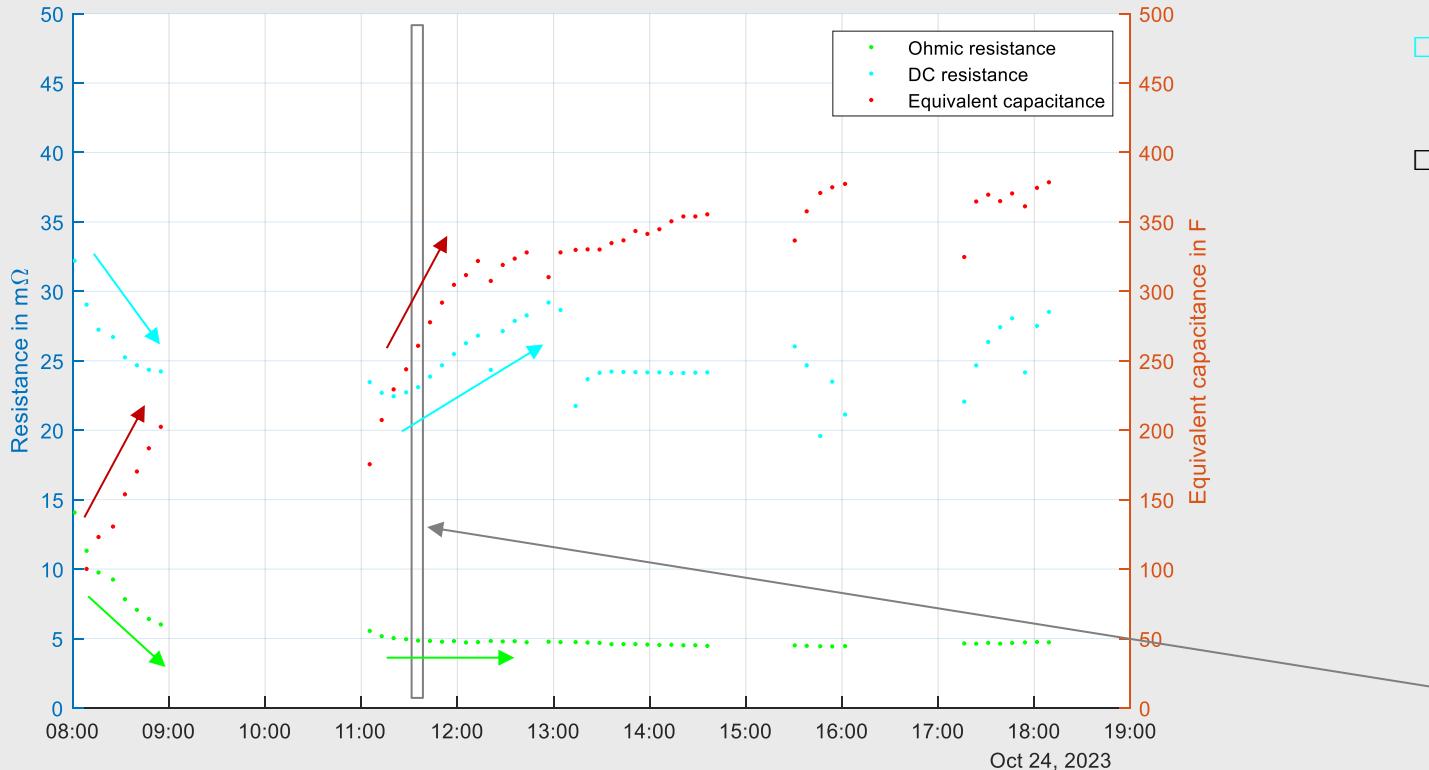


Specification dch_step1:

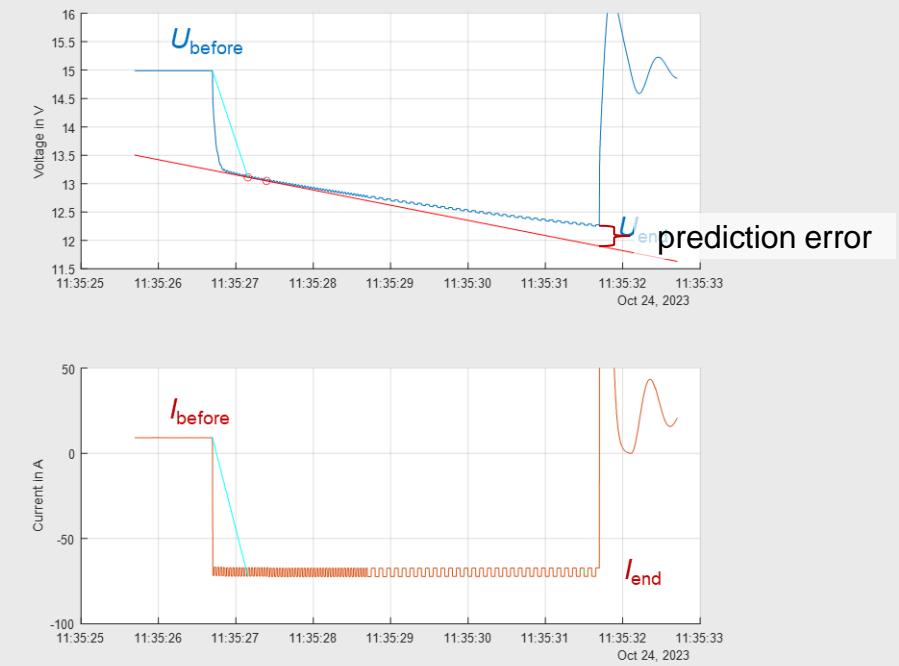
- 70 A for 5+ s*
- 12.0 V voltage limit
- superimposed ripples $\pm 10 \text{ A} / \pm 0.1 \text{ V}$

* duration depending on test bench control

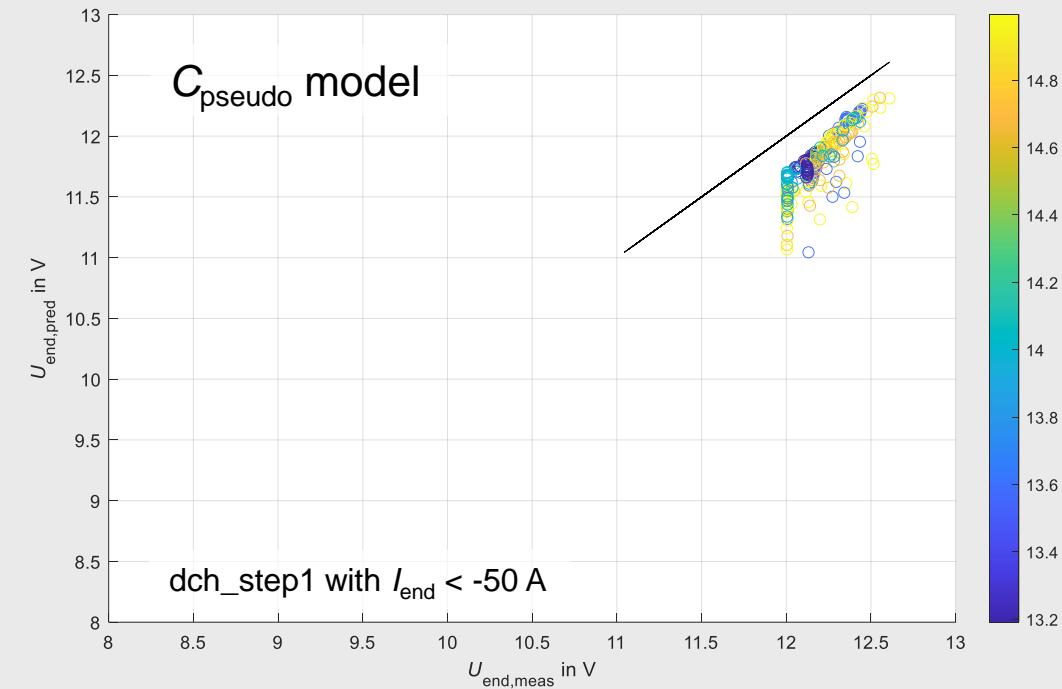
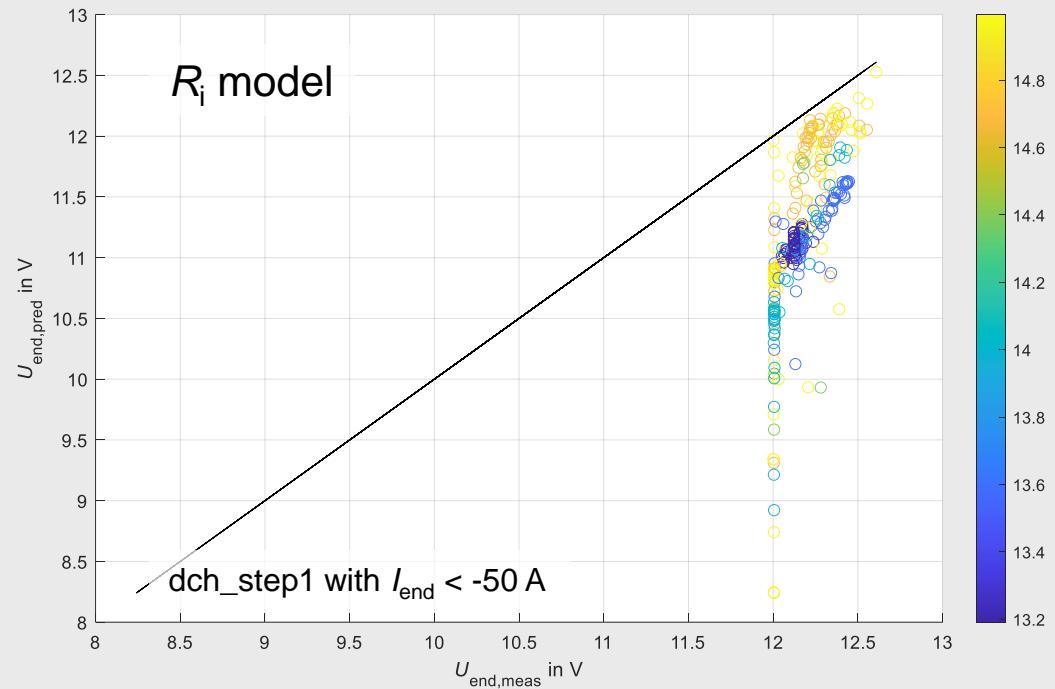
Model 2: Example for parameter evolution during one trip



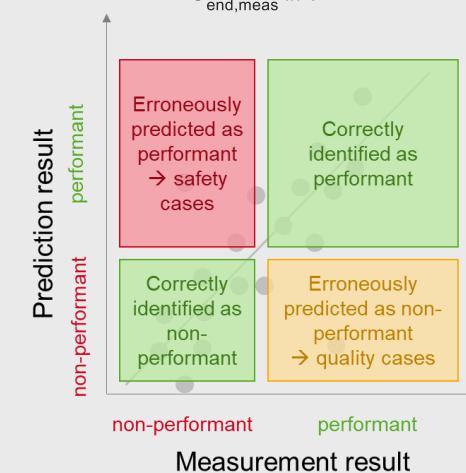
- Example for transient recovery
 - R_{DC} first decreases almost parallel to R_i , but then diverges significantly over time
 - Pseudo capacitance C_{pseudo} builds up in course of the day and contributes significantly to recovery of battery performance



Comparison of model predictions



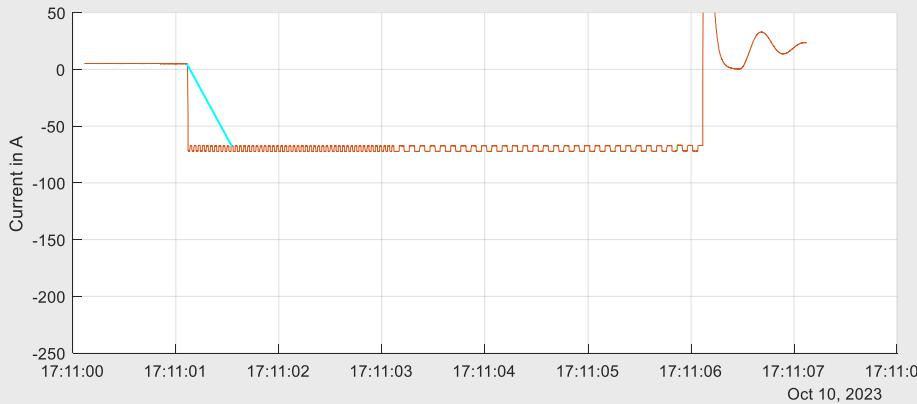
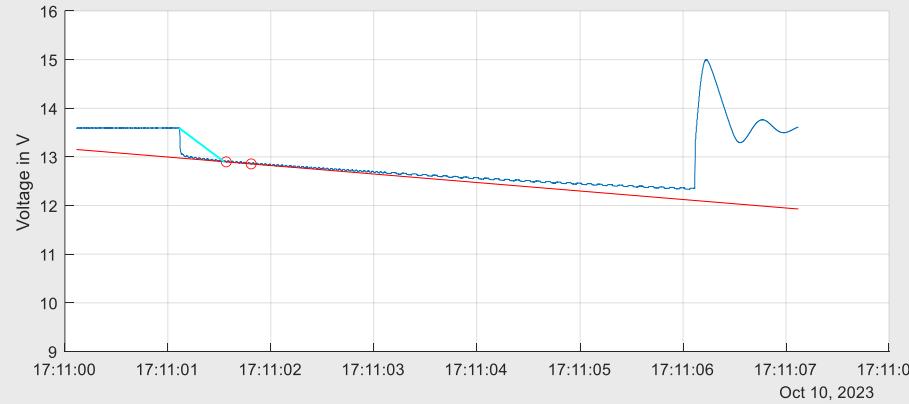
- Both models systematically do not show safety cases
- C_{pseudo} model reduces prediction error vs. R_i model
- Quality cases can be reduced further by model enhancements



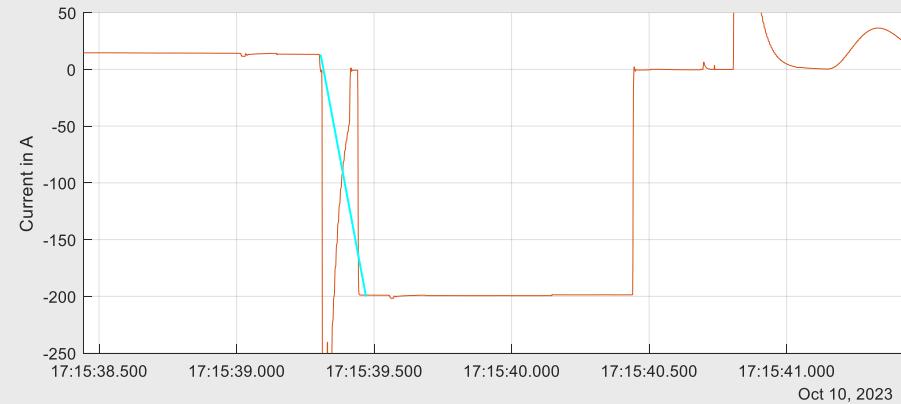
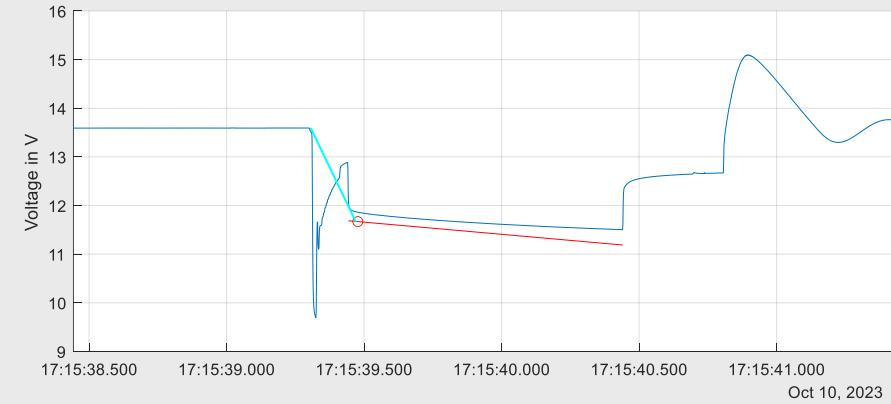
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SSoF0 prediction – low error example

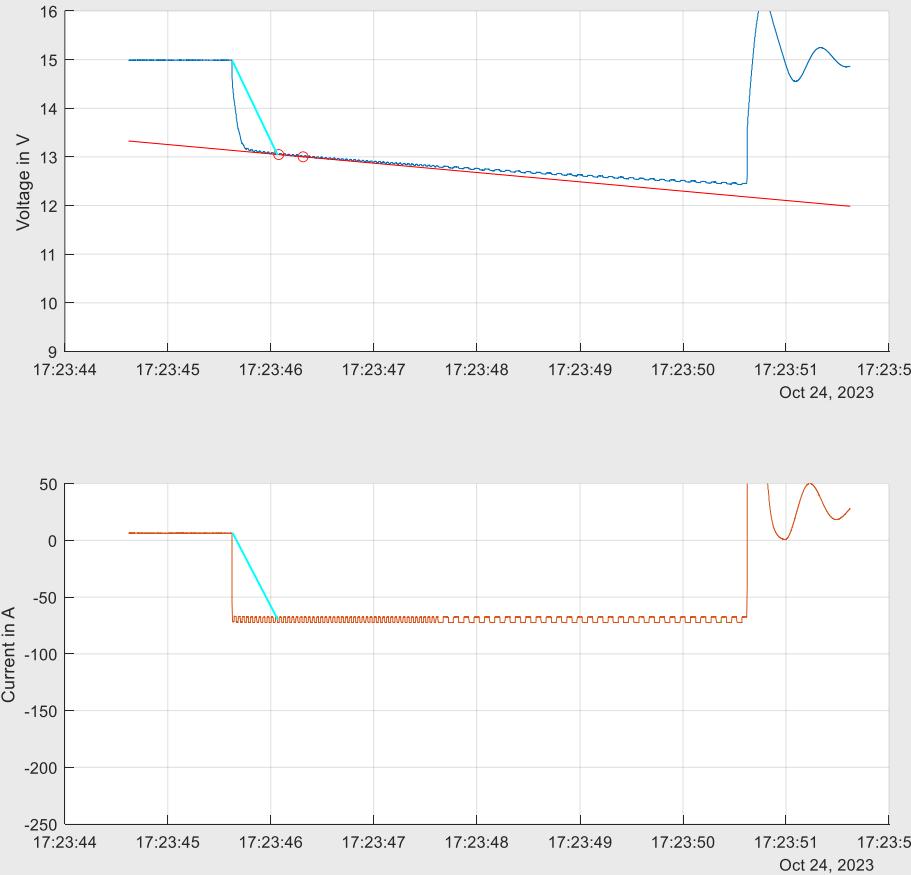


prediction

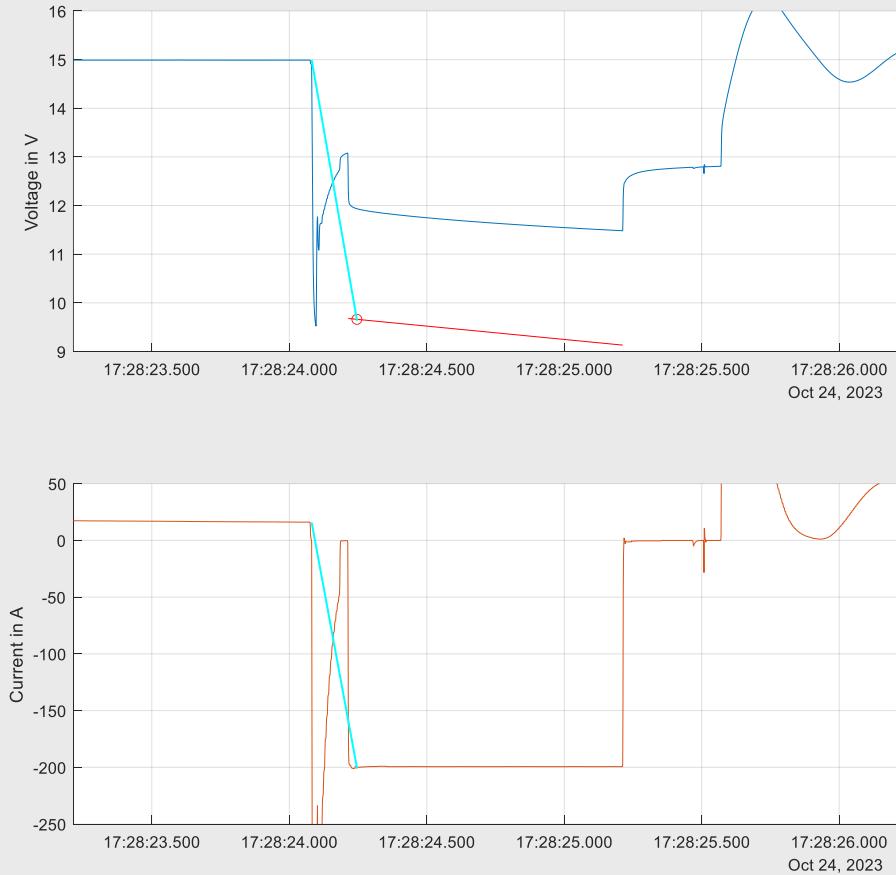


- Good accuracy for initial voltage drop despite extrapolation 70 A to 200 A
- Good accuracy for pseudo-capacitive behavior

SSoF0 prediction – higher error example

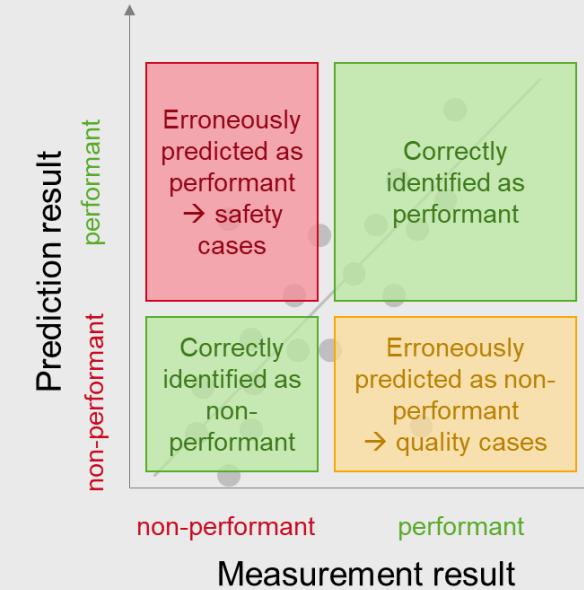
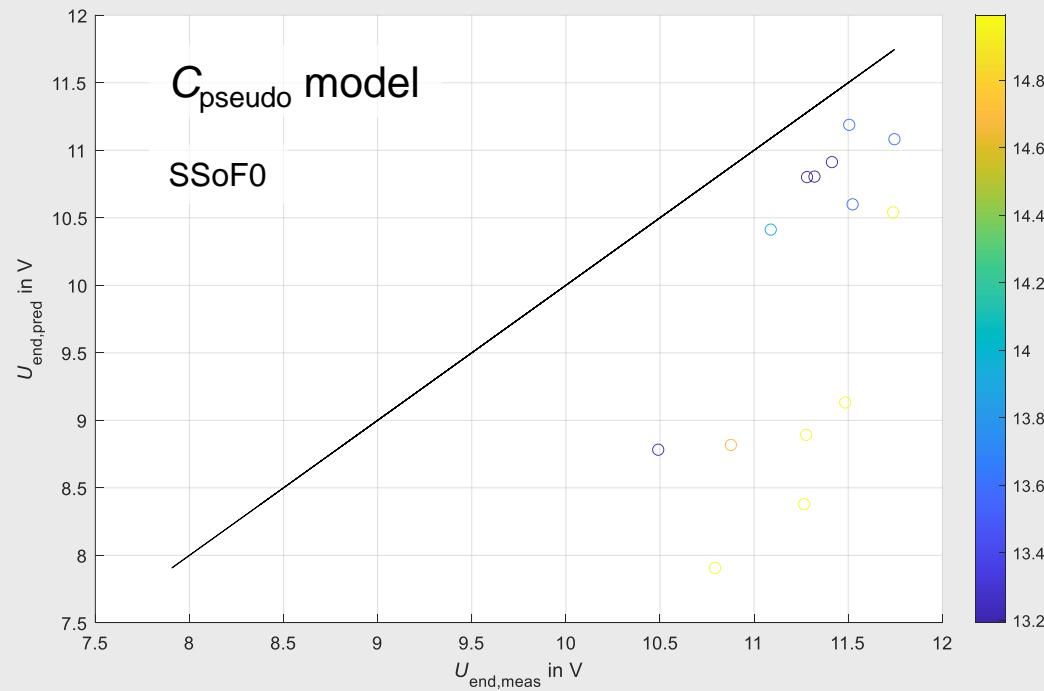


prediction



- Higher error due to charging/discharging linearization: High charging voltage with low charging current
 - 1.5...2.0 V for 6.4 A charging vs. 0.5...1.0 V for 70 A discharging

Model prediction results



- C_{pseudo} model systematically does not show safety cases
 - Good and safe extrapolation from 70 A to 200 A (Butler-Volmer behavior)
- Quality cases can be reduced further by model enhancements
 - Step-wise linear R_{DC} for charging/discharging, elevated discharge voltage plateau

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- CBI SSoF database can be used to analyze transient behavior of starter and auxiliary batteries under realistic operating conditions
 - Only a small fraction of available data used in this study
- This study employs passive equivalent circuits ...
 - ... that do not require parameterization for SoC and temperature
 - ... but adaptation to observed battery behavior in vehicle
- Short-term prediction of discharge peak voltages demonstrated with two primitive models
 - Primary functional safety requirement fulfilled → No false negative prediction
 - Vehicle availability may be compromised by large false positive errors
- Further model improvements could be achieved by
 - Step-wise linear R_{DC} for charging/discharging
 - Modeling elevated discharge voltage plateau (Pilatowicz)

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