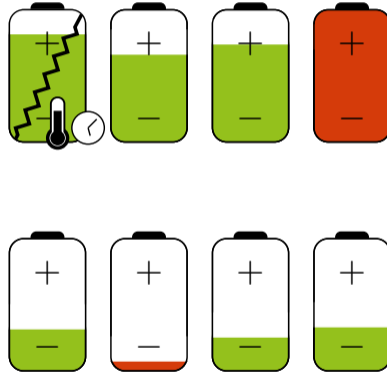


Set-based identification of the open-circuit voltage characteristic of battery cells aiming at detecting aging effects

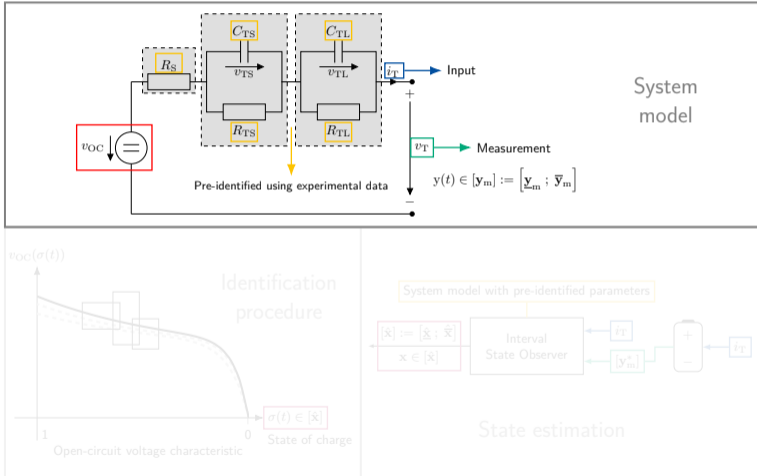
Marit Lahme, October 30, 2024

Carl von Ossietzky Universität Oldenburg, Germany

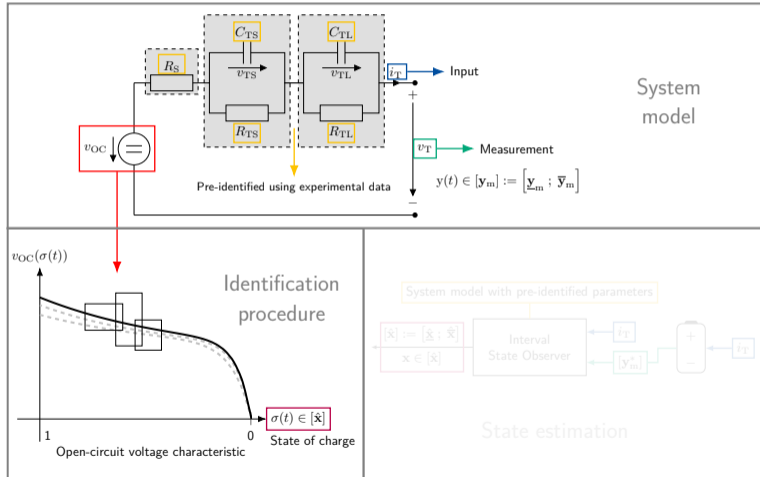
Motivation



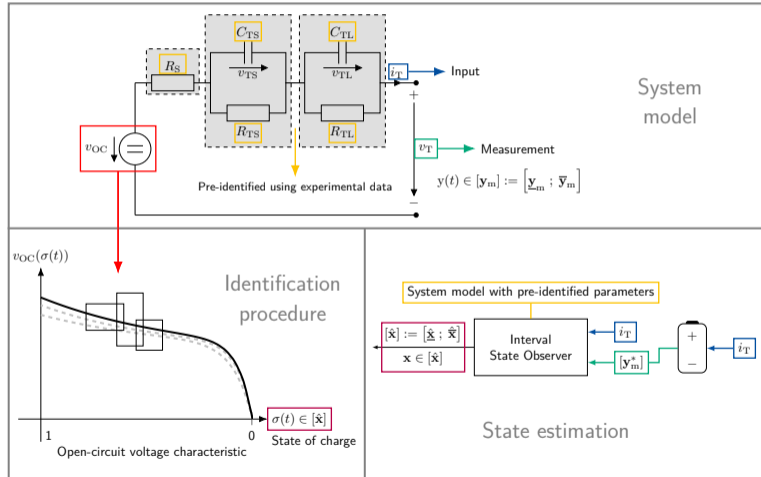
Objective



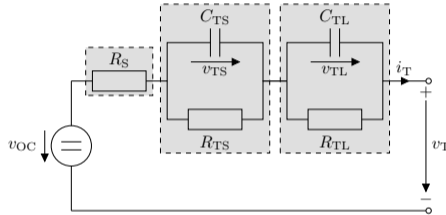
Objective



Objective



Equivalent Circuit Model of a Lithium-Ion Battery Cell



SOC-dependent parameters

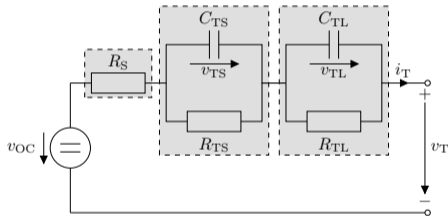
$$R_{\iota}(t) = R_{\iota a} \cdot e^{R_{\iota b} \cdot \sigma(t)} + R_{\iota c}$$

$$C_{\kappa}(t) = C_{\kappa a} \cdot e^{C_{\kappa b} \cdot \sigma(t)} + C_{\kappa c}$$

$$\iota \in \{S, TS, TL\}, \kappa \in \{TS, TL\}$$

$R_{\iota a}, R_{\iota b}, R_{\iota c}, C_{\kappa a}, C_{\kappa b}, C_{\kappa c}$ are identified beforehand based on experimental data.

Equivalent Circuit Model of a Lithium-Ion Battery Cell



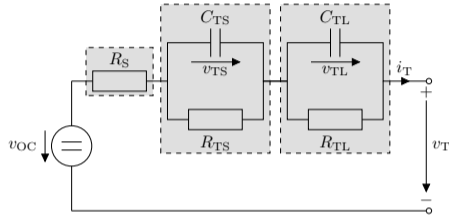
State equations

$$\dot{\sigma}(t) = -\frac{i_T(t)}{C_{\text{Bat}}}$$

$$\dot{v}_\iota(t) = \frac{-v_\iota(t)}{C_\iota(t) \cdot R_\iota(t)} + \frac{i_T(t)}{C_\iota(t)}$$

$$\sigma(t) \in [0 ; 1] , \quad \iota \in \{\text{TS}, \text{TL}\}$$

Equivalent Circuit Model of a Lithium-Ion Battery Cell

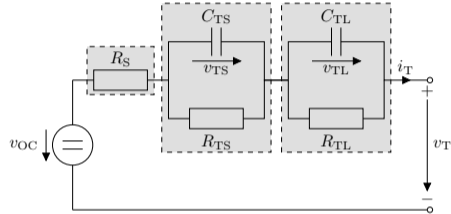


Open-circuit voltage

$$v_{OC}(\sigma(t)) = v_0 \cdot e^{v_1 \cdot \sigma(t)} + v_2 + v_3 \cdot \sigma(t) + v_4 \cdot \sigma(t)^2 + v_5 \cdot \sigma(t)^3$$

$v_0, v_1, v_2, v_3, v_4, v_5$ are identified beforehand based on experimental data.

Equivalent Circuit Model of a Lithium-Ion Battery Cell



Output equation

$$v_T(t) = v_{OC}(\sigma(t)) - v_{TS}(t) - v_{TL}(t) - i_T(t) \cdot R_S(\sigma(t))$$

Equivalent Circuit Model of a Lithium-Ion Battery Cell

Nonlinear expression for the open-circuit voltage

$$v_{OC}(\sigma(t)) = v_0 \cdot e^{v_1 \cdot \sigma(t)} + v_2 + v_3 \cdot \sigma(t) + v_4 \cdot \sigma(t)^2 + v_5 \cdot \sigma(t)^3$$

Quasi-linear expression for the open-circuit voltage

$$\begin{aligned}\tilde{v}_{OC}(\sigma(t)) &= \eta_{OC}(\sigma(t)) \cdot \sigma(t) \\ &= v_{OC}(\sigma(t)) - v_0 - v_2 \\ &= \left(v_0 \cdot \frac{e^{v_1 \cdot \sigma(t)} - 1}{\sigma(t)} + v_3 + v_4 \cdot \sigma(t) + v_5 \cdot \sigma^2(t) \right) \cdot \sigma(t)\end{aligned}$$

Equivalent Circuit Model of a Lithium-Ion Battery Cell

Quasi-linear output equation

$$\mathbf{y}(t) = v_T(t) = \tilde{v}_{OC}(t) - v_{TS}(t) - v_{TL}(t) - i_T(t) \cdot R_S(t)$$

Quasi-linear input-independent output equation

$$\begin{aligned} \mathbf{y}^*(t) &= \mathbf{y}(t) + i_T(t) \cdot R_S(t) \\ &= \tilde{v}_{OC}(t) - v_{TS}(t) - v_{TL}(t) \end{aligned}$$

State-Space Representation

State vector and measurement

$$\mathbf{x}(t) = [\sigma(t) \quad v_{\text{TS}}(t) \quad v_{\text{TL}}(t)]^T$$
$$\mathbf{y}^*(t) = v_{\text{T}}(t) + i_{\text{T}}(t) \cdot R_{\text{S}}(t)$$

Quasi-linear, continuous-time state-space model

$$\dot{\mathbf{x}}(t) = \mathbf{A}(\sigma(t)) \cdot \mathbf{x}(t) + \mathbf{b}(\sigma(t)) \cdot i_{\text{T}}(t)$$
$$\mathbf{y}^*(t) = \mathbf{c}^T(\sigma(t)) \cdot \mathbf{x}(t)$$

Purpose of the quasi-linear representation: to make linear state observers applicable.

State-Space Representation

$$\mathbf{A}(\sigma(t))$$

$$\begin{bmatrix} 0 & 0 & 0 \\ 0 & \frac{-1}{C_{\text{TS}}(\sigma(t)) \cdot R_{\text{TS}}(\sigma(t))} & 0 \\ 0 & 0 & \frac{-1}{C_{\text{TL}}(\sigma(t)) \cdot R_{\text{TL}}(\sigma(t))} \end{bmatrix}$$

$$\mathbf{b}(\sigma(t))$$

$$\begin{bmatrix} \frac{-1}{C_{\text{Bat}}} \\ \frac{1}{C_{\text{TS}}(\sigma(t))} \\ \frac{1}{C_{\text{TL}}(\sigma(t))} \end{bmatrix}$$

$$\mathbf{c}^T(\sigma(t))$$

$$[\eta_{\text{OC}}(\sigma(t)) \quad -1 \quad -1]$$

Interval Observer

Goal	Bound the true value, so that $\underline{\hat{x}} \leq \mathbf{x} \leq \hat{\bar{x}}$
Prerequisite	choose initial condition so that $\underline{\hat{x}}_0 \leq \mathbf{x}_0 \leq \hat{\bar{x}}_0$ observer system matrix has a Metzler structure (monotonicity condition)

Metzler matrix

All off-diagonal elements are non-negative.

Interval Observer

This interval observer is based on two Luenberger observers.

- one observer to estimate the upper bound
- one observer to estimate the lower bound

Example of a Luenberger observer

$$\begin{aligned}\dot{\hat{\mathbf{x}}} &= \mathbf{A} \cdot \hat{\mathbf{x}} + \mathbf{b} \cdot i_T + \mathbf{H} \cdot (y_m - \hat{y}) \\ &= \mathbf{A} \cdot \hat{\mathbf{x}} + \mathbf{b} \cdot i_T + \mathbf{H} \cdot (y_m - \mathbf{c}^T \cdot \hat{\mathbf{x}}) \\ &= (\mathbf{A} - \mathbf{H} \cdot \mathbf{c}^T) \cdot \hat{\mathbf{x}} + \mathbf{b} \cdot i_T + \mathbf{H} \cdot y_m\end{aligned}$$

Interval Observer

$$\mathbf{A}(\sigma(t)) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -|a_{22}| & 0 \\ 0 & 0 & -|a_{33}| \end{bmatrix}, \mathbf{c}^T(\sigma(t)) = [+|c_{11}| \quad -1 \quad -1], \mathbf{H} = \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix}$$

$$\mathbf{A} - \mathbf{H} \cdot \mathbf{c}^T = \begin{bmatrix} -h_1 \cdot |c_{11}| & h_1 & h_1 \\ -h_2 \cdot |c_{11}| & -|a_{22}| + h_2 & h_2 \\ -h_3 \cdot |c_{11}| & h_3 & -|a_{33}| + h_3 \end{bmatrix} \hat{=} \begin{bmatrix} * & \geq 0 & \geq 0 \\ \geq 0 & * & \geq 0 \\ \geq 0 & \geq 0 & * \end{bmatrix}$$

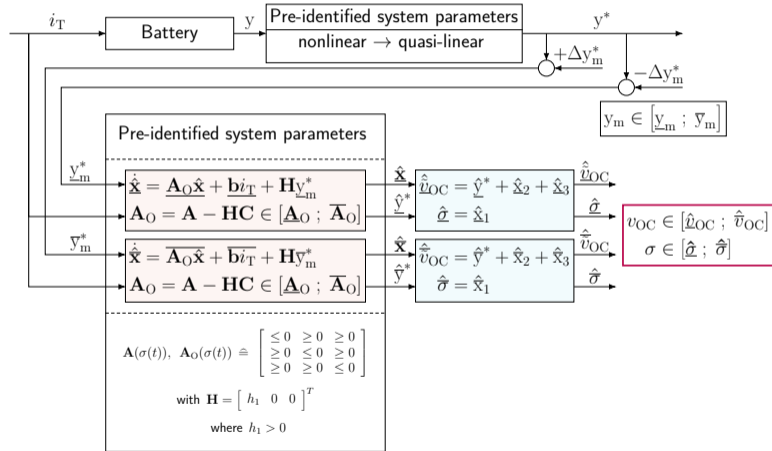
Interval Observer

$$\mathbf{A}(\sigma(t)) = \begin{bmatrix} 0 & 0 & 0 \\ 0 & -|a_{22}| & 0 \\ 0 & 0 & -|a_{33}| \end{bmatrix}, \mathbf{c}^T(\sigma(t)) = [+|c_{11}| \quad -1 \quad -1], \mathbf{H} = \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix}$$

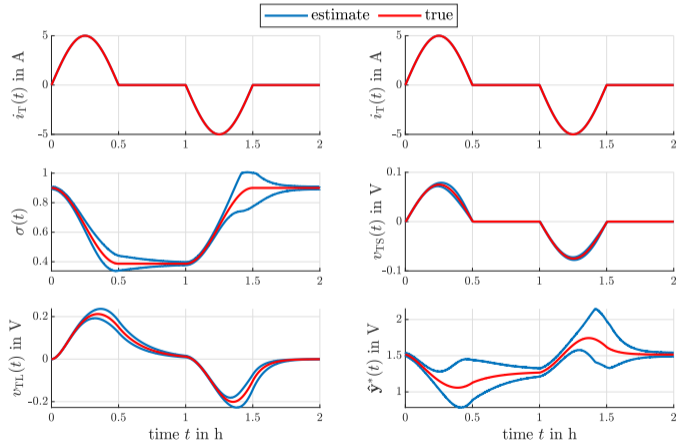
$$\mathbf{A} - \mathbf{H} \cdot \mathbf{c}^T = \begin{bmatrix} -h_1 \cdot |c_{11}| & h_1 & h_1 \\ -h_2 \cdot |c_{11}| & -|a_{22}| + h_2 & h_2 \\ -h_3 \cdot |c_{11}| & h_3 & -|a_{33}| + h_3 \end{bmatrix} \hat{=} \begin{bmatrix} * & \geq 0 & \geq 0 \\ \geq 0 & * & \geq 0 \\ \geq 0 & \geq 0 & * \end{bmatrix}$$

$$\mathbf{H} = [h_1 \quad 0 \quad 0]^T \text{ with } h_1 > 0$$

Interval Observer-Based State Estimation



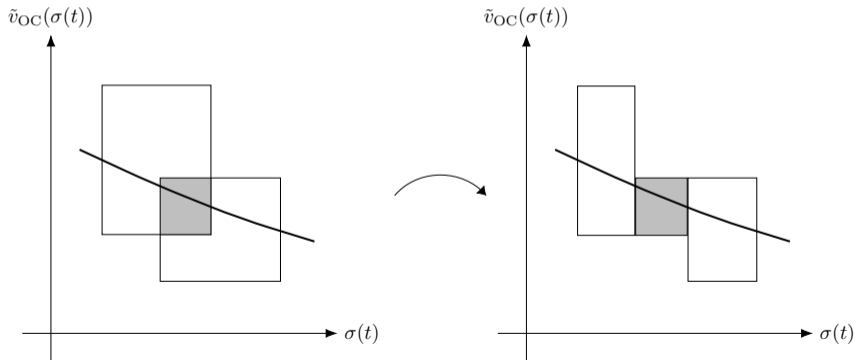
Interval Observer-Based State Estimation



Identification procedure

Estimation result based on the interval observer

$\forall \sigma_j \in [\sigma_j], \exists \tilde{v}_{OC,j} \in [\tilde{v}_{OC,j}]$ s.t. $\tilde{v}_{OC,j} = \tilde{v}_{OC}(\sigma_j)$, $j = 1, \dots,$



Simulation result

