Structural Identifiability of a Pseudo-2D Li-ion Battery Electrochemical Model

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Abstract: Growing demand for fast charging and optimised battery designs is fuelling significant interest in electrochemical models of Li-ion batteries. However, estimating parameter values for these models remains a major challenge. In this paper, a structural identifiability analysis was applied to a pseudo-2D Li-ion electrochemical battery model that can be considered as a linearised and decoupled form of the benchmark Doyle-Fuller-Newman model. From an inspection of the impedance function, it was shown that this model is uniquely parametrised by 21 parameters, being combinations of the electrochemical parameters like the conductivities and diffusion coefficients. The well-posedness of the parameter estimation problem with these parameters was then established. This result could lead to more realistic predictions about the internal state of the battery by identifying the parameter set that can be uniquely identified from the data.

Keywords: Li-ion batteries, electrochemical models, structural identifiability.

INTRODUCTION

As the importance of electrical energy storage continues to grow and Li-ion batteries numbers keep on increasing, this technology is becoming ever more mature. The sheer number of Li-ion batteries being produced is leading to significant price reductions (Curry (2017)), due primarily to the economics of scale, meaning that any new battery chemistry hoping to be commercialised will have to overcome increasingly steep economic barriers before entering the mass-market. As such, at least in the near future, the next wave of battery innovation is likely to come from optimising existing technology, rather than from the introduction of radically new chemistries. Control engineering will play a pivotal role in this optimisation.

Electrochemical models are important tools for optimising battery use, both for control purposes within a battery management system (Chaturvedi et al. (2010)) and for design. These models provide a rich description of the battery’s response in terms of snapshots of its internal electrochemical state; however, their accuracy critically relies upon the electrochemical parameter values. Obtaining accurate estimates of these parameters has thus emerged as a crucial research topic in recent years. Unfortunately, the relative complexity of electrochemical models makes estimating their parameter values difficult. This difficulty was illustrated in Forman et al. (2012) where it took approximately three weeks’ worth of computation on five quad-core Intel Q8200 computers to estimate the parameters of the benchmark Doyle-Fuller-Newman (DFN) model (Doyle et al. (1993)). To bring some clarity to this problem, Bizeray et al. (2018) recently approached the parameter estimation problem from a different perspective by performing a structural identifiability analysis of the simplified single particle model (SPM), determining the unique six parameters fully describing the SPM’s response. In such a way, the well-posedness of the parameter estimation problem for this simplified model could be established. This result built upon existing literature in this area, including Pozzi et al. (2018) where sensitivity functions for a polynomial approximations of the single particle model with electrolyte (SPMe) were obtained, Park et al. (2018) which considered experiment design for ranking the sensitivity of each electrochemical parameter, Lai et al. (2019) where analytic sensitivity functions for the SPM’s parameters were derived from Padé approximations of the impedance and Alavi et al. (2016) that analysed the structural identifiability of Randles circuit battery models.

Motivated by Bizeray et al. (2018), this paper develops a framework towards a structural identifiability analysis of the DFN electrochemical Li-ion battery model (Doyle et al. (1993)). The DFN model is widely considered as a benchmark micro-scale model for Li-ion batteries from which simplifications like the SPM (Ning and Popov (2004); Atlung et al. (1979)) and the SPMe (Moura et al. (2017); Marquis et al. (2019); Richardson et al. (2019)) are derived. But, estimating its parameter values in a methodical way is challenging (Forman et al. (2012)). This paper details preliminary results on this problem by performing a structural identifiability analysis of a decoupled and linearised form of the DFN model which can be considered as the SPMe with added double-layer effects.

To perform the structural identifiability analysis, the DFN model was first simplified into a linearised form with dynamics decomposed into three elements: solid-state dif-
fusión, bulk electrolyte mobility and the charge transfer resistance caused by the relaxation of the overpotential. This simplification enables a tractable analysis, with the generalised impedance functions developed in Sikka and White (2007) being too complex, right now, to determine structural identifiability. It is shown that this decoupled DFN model is uniquely parametrised from current/voltage data by 21 parameters formed from combinations of the electrochemical parameters (see Table 2). It is hoped that the results of this work will provide the theoretical underpinning behind a generalised parameter estimation method of the DFN model that will not rely upon substantial a priori knowledge of the cell’s makeup. The need for accurate, and recursive, estimates of the DFN model’s parameters is expected to grow still further as the importance of fast charging (Kang and Ceder (2009)) and battery design become ever clearer. For these applications, the SPM provides neither sufficient accuracy nor richness.

NOTATION

Definitions of the electrochemical model states and parameters are given in Table 1. Special attention is paid to the definition of the spatial domains, identified by the subscript \( k \in \{1, 2, 3\} \), \( k = 1 \) denoting the anode, \( k = 2 \) the separator and \( k = 3 \) the cathode. For instance, the point \( x = L_{12} \) defines the anode/separator boundary and \( x = L_{23} \) is the separator/cathode boundary. \( \Omega_1 := \{x : 0 < x < L_{12}\} \) denotes the spatial domain of the anode, \( \Omega_2 := \{x : L_{12} < x < L_{23}\} \) that of the separator and \( \Omega_3 := \{x : L_{23} < x < L\} \) that of the cathode. When a parameter takes different values in each domain then the subscript \( k \) is added, for instance \( \varepsilon_1 \) is the porosity in the anode. When these parameter subscripts are missing, like in (6), this indicates that the equation holds in all domains. This notation is adopted for compactness to avoid duplication of the equations as much as possible.

For spatio-temporal variables (e.g. \( c_e(x, t) \)), tilde will denote variations (e.g. \( \tilde{c}_e(x, t) \)) around an equilibrium denoted by an \( \ast \) (e.g. \( c_e^\ast \)). In particular, the expansions

\[
\begin{align*}
  c_e(x, t) &= c_e^\ast + \tilde{c}_e(x, t), \\
  u_s(x, r, t) &= \tilde{u}_s(x, r, t) + \frac{R_s}{R_s} \tilde{c}_e^{\text{surf}}(x, t), \\
  u_s(x, r, t) &= \kappa c_e(x, r, t)
\end{align*}
\]

are used. The surface concentrations are \( c_e^{\text{surf}}(x, t) = c_e(x, R_s, t) \) and the transformed surface concentrations are \( u_e^{\text{surf}}(x, t) = u_e(x, R_s, t) \). Overbars denote Laplace transforms of signals, e.g. \( \bar{c}_e(x, s) \).

1. PROBLEM FORMULATION

Structural identifiability of a decomposed form of the DFN electrochemical model for Li-ion batteries is considered. In each electrode, the DFN model electrochemical equations are

\[
\begin{align*}
  \frac{\partial u_s(x, r, t)}{\partial t} &= D_s \frac{\partial^2 u_s(x, r, t)}{\partial r^2}, \\
  \frac{\partial c_e(x, t)}{\partial t} &= D_a \frac{\partial^2 c_e(x, t)}{\partial x^2} + a_s(1 - \varepsilon) j(x, t), \\
  \eta(x, t) &= \phi_s(x, t) - \phi_e(x, t) - U(u_e^{\text{surf}}(x, t)), \\
  i_0 &= K F(c_e^{\text{max}} - c_e^{\text{surf}}(x, t))^\alpha c_e^{\text{surf}}(x, t)^{\alpha - 1}, \\
  j(x, t) &= \frac{i_0(x, t)}{F} \sinh \left( \frac{F}{RT} \eta(x, t) \right), \\
  \frac{\partial i_e(x, t)}{\partial x} &= a_s F j(x, t), \\
  \frac{\partial c_e(x, t)}{\partial t} &= D_s \frac{\partial^2 c_e(x, t)}{\partial x^2}, \\
  \frac{\partial i_i(x, t)}{\partial x} &= -\frac{\partial \phi_s(x, t)}{\partial x} + K \frac{\partial c_e(x, t)}{\partial x}, \\
  \frac{\partial \eta(x, t)}{\partial t} &= \frac{i_0(x, t)}{F} \sinh \left( \frac{F}{RT} \eta(x, t) \right)
\end{align*}
\]
Table 1. Nomenclature of the DFN model.

| Sub and super scripts | Parameters for solid diffusion | Parameters for bulk electrolyte | Parameters for overpotential |
|-----------------------|--------------------------------|--------------------------------|------------------------------|
| \(1\) Denotes anode parameters. | \(\tau^c_{e,k} \quad \frac{D_{e,k}}{\varepsilon_F R_s^{\varepsilon_F}} \quad \text{Particle time constant.} \) | \(\tau^e_{e,k} \quad \frac{D_{e,k}}{\varepsilon_F R_s^{\varepsilon_F}} \quad \text{Electrolyte time constant.} \) | \(\tau^e_{\eta,k} \quad \varepsilon_F R_s^{\varepsilon_F} \quad \text{Double layer time constant.} \) |
| \(2\) Denotes separator parameters. | \(\theta^c_{e,k} \quad \frac{\partial^2 \tilde{E}_e(t)}{\partial x^2} = \frac{a_s(1 - t_{+})}{\varepsilon F} j^*(t) \), (6a) | \(\theta^e_{e,k} \quad \frac{\partial^2 \tilde{u}_e(t)}{\partial x^2} \), (6b) | \(\theta^e_{\eta,k} \quad \varepsilon_F R_s^{\varepsilon_F} \quad \text{Normalised reactivity term.} \) |
| \(3\) Denotes cathode parameters. | \(\tilde{E}_e(x,t) = \frac{\partial^2 \tilde{E}_e(x,t)}{\partial x^2} - \frac{a_s(1 - t_{+})}{\varepsilon F} j^*(t) \), (6a) | \(\tilde{u}_e(x,t) = \frac{\partial^2 \tilde{u}_e(x,t)}{\partial x^2} \), (6b) | \(\tilde{u}_e(x,0,t) = 0 \), (7b) |
| \(12\) Anode/seperator boundary. | \(\frac{1}{R_s} \frac{\partial \tilde{u}_e(x,t)}{\partial r} \bigg|_{r=R_s} \tilde{u}_e(x,R_s,t) = -j^*(t) \), (7a) | \(\partial \eta(x,t) \bigg|_{x \in (0,L)} = -\frac{i(t)}{\sigma} \), (8a) | \(\frac{\partial \eta(x,t)}{\partial x} \bigg|_{x \in \{12,23\}} = -\frac{i(t)}{\kappa} \), (8b) |
| \(23\) Cathode/seperator boundary. | \(\tilde{u}_e(x,0,t) = 0 \), (7b) | Each of these dynamical equations will be analysed individually to establish structural identifiability. The electrochemical states of this system are subject to the boundary conditions | at the time scales where these dynamics are important (doing so decouples the dynamics of \(\partial dt \) from \(c_e \)) and by adding a double layer term to the divergence equation (2)). | The obtained linearised state-space form of (2) is then (Drummond et al. (2019)) |
\[ \frac{\partial \bar{c}_e(x,t)}{\partial x} \bigg|_{x \in \{0,L\}} = 0, \quad (9a) \]
\[ D_{c,1} \frac{\partial c_e}{\partial x} \bigg|_{an} = D_{c,2} \frac{\partial c_e}{\partial x} \bigg|_{sep}, \quad x = L_{12}, \]
\[ D_{c,1} \frac{\partial c_e}{\partial x} \bigg|_{cat} = D_{c,2} \frac{\partial c_e}{\partial x} \bigg|_{sep}, \quad x = L_{23}, \]
\[ \bar{c}_e(x,t) \text{ is continuous on } x \in \{ L_{12}, L_{23} \}. \]

Here, the subscript \( an \) denotes a derivative normal to the boundary of the anode, \( sep \) is normal to the separator and \( cat \) is normal to the cathode.

### 1.1 Voltage expression

The voltage is defined as the difference in solid-phase potentials between each current collector but requires some reformulation to be expressed in terms of the electrochemical states. For this, consider

\[ v(t) = \phi_a(L,t) - \phi_a(0,t) - R_{ctc}(t) \]

where

\[ v(t) = \phi_a(L,t) - \phi_a(0,t) + \phi_e(L,t) - \phi_e(0,t) - R_{ctc}(t) \]

\[ = \phi_a(L,t) - \phi_a(0,t) + \int_0^L \frac{\partial \phi_a(x,t)}{\partial x} \, dx - R_{ctc}(t) \]

with \( \frac{\partial \phi_a}{\partial x} \) evaluated from Machinnes’ equation (2g)

\[ v(t) = \frac{\partial \phi_a(L,t)}{\partial x} - \frac{\partial \phi_a(0,t)}{\partial x} + \int_0^L \frac{\partial \phi_a(x,t)}{\partial x} \, dx - R_{ctc}(t). \]

Analysing the integral in (10d) piece-wise across each of the three domains \( \Omega_1, \Omega_2, \Omega_3 \) using the relationship

\[ i_e(x,t) = \frac{\sigma_1 \kappa_k}{\sigma_k + \kappa_k} \frac{\partial \phi_a}{\partial x} + \frac{K \sigma_1 \kappa_k}{\sigma_k + \kappa_k} \frac{\partial \bar{c}_e}{\partial x} + \frac{\kappa_k i(t)}{\sigma_k + \kappa_k} \]

\[ i_e(x,t) = i(t), \quad x \in \Omega_2, \]

allows the voltage to be expressed in terms of the electrochemical states

\[ v(t) = \phi_a(L,t) - \phi_a(0,t) \]

\[ + \int_{\Omega_1} - \frac{\sigma_1}{\sigma_1 + \kappa_k} \frac{\partial \phi_a}{\partial x} + \frac{K \kappa_1}{\sigma_1 + \kappa_1} \frac{\partial \bar{c}_e}{\partial x} \, dx - \frac{i(t)}{\sigma_1 + \kappa_1} \]

\[ + \int_{\Omega_3} - \frac{\sigma_3}{\sigma_3 + \kappa_3} \frac{\partial \phi_a}{\partial x} + \frac{K \kappa_3}{\sigma_3 + \kappa_3} \frac{\partial \bar{c}_e}{\partial x} \, dx - \frac{i(t)}{\sigma_3 + \kappa_3} \]

\[ + \int_{\Omega_3} K \frac{\partial \bar{c}_e}{\partial x} \frac{i(t)}{\kappa_k} \, dx - R_{ctc}(t). \]

However, in this paper, the analysis is concerned with local perturbations of the voltage around the equilibrium of the open circuit voltage \( (OCV) \). These local perturbations are found by substituting the expression for \( \phi_a \) from the overpotential definition (2c) into (12) and using the approximation

\[ U(\bar{c}_e^{surf}(x,t), \sigma^{surf}, \sigma^*) \approx U(\bar{c}_e^{surf}, \sigma^{surf}) + \frac{1}{R_s} \frac{dU(\bar{c}_e^{surf})}{d\bar{c}_e^{surf}} |_{\bar{c}_e^{surf} = \bar{c}_e^{surf}, \sigma^{surf}, \sigma^*} \tilde{u}_s^{surf}. \]

The voltage can then be expressed as

\[ v(t) = OCV + \tilde{v}(t), \]

with the OCV defined from the \( U(\bar{c}_e^{surf}, \sigma^*) \) terms and the perturbations being

\[ \tilde{v}(t) = v_y(t) + \tilde{v}_{usurf}(t) + \tilde{v}_{ce}(t) - R_{res}(t) \]

where

\[ v_y(t) = - \frac{\kappa_1}{\kappa_1 + \sigma_1} \tilde{u}_s(L_{12}, t) + \frac{\kappa_2}{\kappa_2 + \sigma_2} \tilde{u}_s(L_{23}, t) + \frac{\kappa_3}{\kappa_3 + \sigma_3} \tilde{u}_s(L_{12}, t) \]

\[ + \frac{\sigma_1}{\sigma_1 + \kappa_1} \frac{d\tilde{u}_s}{dt} \]

\[ + \frac{\sigma_2}{\sigma_2 + \kappa_2} \frac{d\tilde{u}_s}{dt} \]

\[ + \frac{\sigma_3}{\sigma_3 + \kappa_3} \frac{d\tilde{u}_s}{dt}, \]

\[ \tilde{v}_{usurf}(t) = - \frac{K \sigma_1}{\sigma_1 + \kappa_1} \tilde{u}_s^{surf}(0, t) + \frac{K \sigma_2}{\sigma_2 + \kappa_2} \tilde{u}_s^{surf}(L_{23}, t) + \frac{K \sigma_3}{\sigma_3 + \kappa_3} \tilde{u}_s^{surf}(L_{12}, t) \]

\[ + \frac{\kappa_1}{\kappa_1 + \sigma_1} \tilde{u}_s^{surf}(L_{12}, t), \]

\[ \tilde{v}_{ce}(t) = - \frac{K \kappa_1}{\kappa_1 + \sigma_1} \tilde{c}_e(L_{12}, t) + \frac{K \kappa_2}{\kappa_2 + \sigma_2} \tilde{c}_e(L_{23}, t) + \frac{K \kappa_3}{\kappa_3 + \sigma_3} \tilde{c}_e(L_{12}, t), \]

\[ R_{res} = R_{ctc} + \frac{L_1}{\sigma_1 + \kappa_1} + \frac{L_2}{\sigma_2 + \kappa_2} + \frac{L_3}{\sigma_3 + \kappa_3}, \]

Under the constant reactance assumption of the SPM (5), which holds for cells with sufficiently high electronic conductivities, then the voltage from the particles’ surfaces can be simplified to

\[ v_{usurf} \approx - \frac{1}{R_{es,3}} \frac{d\tilde{u}_s^{surf}}{dt} \tilde{u}_s^{surf}(L_{23}, t) - \frac{1}{R_{es,1}} \frac{d\tilde{u}_s^{surf}}{dt} \tilde{u}_s^{surf}(L_{12}, t), \]

allowing the particle dynamics to be described by a single particle.

### 1.2 Structural Identifiability

This paper is concerned with structural identifiability of this system (Bellman and Aström (1970)). Structural identifiability is defined as (Ljung (2001); Alavi et al. (2016)):

**Definition 1.** (Structural identifiability). Consider a model structure \( M \) with the transfer function \( H(s, \theta) \) parametrised by \( \theta \in D \subset \mathbb{R}^n \) where \( n \) denotes the number of parameters of the model. The identifiability equation for \( M \) is given by:

\[ H(s, \theta) = H(s, \theta^*) \]

for almost all \( s \in D \) where \( \theta, \theta^* \in D \). The model structure \( M \) is said to be

- globally identifiable if (17) has a unique solution in \( D \),
- locally identifiable if (17) has a finite number of solutions in \( D \),
- unidentifiable if (17) has an infinite number of solutions in \( D \).

If the model is globally identifiable then it is said to be structurally identifiable and the parameter estimation problem is well-posed (meaning that it admits a unique solution). To verify structural identifiability of the considered model, the minimum number of parameters needed to uniquely characterise the impedance functions of each electrochemical state is determined.
2. IMPEDANCE ANALYSIS FOR STRUCTURAL IDENTIFIABILITY

The analysis of the previous section, under the given assumptions, decomposed the DFN model into three components with three distinct timescales: solid-state diffusion of \( u_s \) (slow); bulk electrolyte diffusion (mid) and relaxation of overpotentials (fast). Structural identifiability of each of the three decoupled timescales will now be analysed individually.

2.1 Low-frequency range: Solid-state diffusion

Solid-state diffusion within the active material particles dominates the low frequency response (when \( \omega \ll 1 \) rad \( s^{-1} \)), with the structural identifiability of these dynamics detailed in Bizeray et al. (2018) and repeated here for completeness. These dynamics are defined by the radial diffusion of (6b), (7), (16) having Laplace transform

\[
\hat{c}_{ic, j}(x, s) = \left( \frac{R^2}{u_s, FD_s} \right) \left( \frac{\tanh \left( \frac{x}{\tau_{cs}} \right)}{\tanh \left( \frac{x}{\tau_{ci}} \right) - \sqrt{\frac{x}{\tau_{ci}}} \tanh \left( \frac{x}{\tau_{si}} \right)} \right).
\]

Applying the constant reactivity approximation of (5) to (18) gives the transfer function of the SPM

\[
\hat{e}_{ic, j}(x, s) = \frac{\theta_{cs, 3} \tau_{cs, 3}}{I(s)} \left( \frac{\tanh \left( \frac{x}{\tau_{cs, 3}} \right)}{\tanh \left( \frac{x}{\tau_{ci, 3}} \right) - \sqrt{\frac{x}{\tau_{si, 3}}} \tanh \left( \frac{x}{\tau_{si, 3}} \right)} \right) - \frac{\theta_{cs, 1} \tau_{cs, 1}}{I(s)} \left( \frac{\tanh \left( \frac{x}{\tau_{cs, 1}} \right)}{\tanh \left( \frac{x}{\tau_{ci, 1}} \right) - \sqrt{\frac{x}{\tau_{si, 1}}} \tanh \left( \frac{x}{\tau_{si, 1}} \right)} \right)
\]

with \( \theta^k_{e, j} = \frac{\partial D_e}{\partial e^k} \) and \( \tau^k_e = R_{e,k} / \sqrt{D_{e,k}} \) where \( k \in \{1, 3\} \). These are the four parameters that uniquely map the input current to variations in particle surface concentrations, under the above assumptions.

2.2 Mid-frequency range: Movement of bulk electrolyte

Moving up the frequency range to \( \omega \approx 1 \) rad \( s^{-1} \), polarisation of the bulk electrolyte becomes the dominant electrochemical effect. Structural identifiability of the electrolyte requires a transfer function for (6a) and (15d). Unfortunately, due to the boundary conditions (9a)-(9d) coupling the electrolyte’s solutions across various domains, the impedance analysis of this state is more complex to analyse than the other effects considered in the model. To alleviate some of this complexity, the uniform reactivity approximation (4) is applied to (6a), converting the dynamics in the electrodes to

\[
\begin{align*}
\frac{\partial \tilde{c}_e(x, t)}{\partial t} &= D_{e, 1} \frac{\partial^2 \tilde{c}_e(x, t)}{\partial x^2} + \pi_1 i(t), \forall x \in \Omega_1, \\
\frac{\partial \tilde{c}_e(x, t)}{\partial t} &= D_{e, 2} \frac{\partial^2 \tilde{c}_e(x, t)}{\partial x^2}, \forall x \in \Omega_2, \\
\frac{\partial \tilde{c}_e(x, t)}{\partial t} &= D_{e, 3} \frac{\partial^2 \tilde{c}_e(x, t)}{\partial x^2} - \pi_3 i(t), \forall x \in \Omega_3,
\end{align*}
\]

where \( \pi_k = \frac{a_{e,k}(1-t_x)}{\varepsilon_x} \), \( k \in \{1, 3\} \) is a parameter for the reactivity.

The co-ordinate transformation \( \gamma \) is introduced, being defined in each electrode as

\[
\begin{align*}
\gamma(x, t) &= \tilde{c}_e(x, t) - \pi_1 \int_0^t i(\tau) d\tau, \quad x \in \Omega_1, \\
\gamma(x, t) &= \tilde{c}_e(x, t) + \pi_3 \int_0^t i(\tau) d\tau, \quad x \in \Omega_3.
\end{align*}
\]

Then, in the electrodes,

\[
\begin{align*}
\frac{\partial \gamma(x, t)}{\partial t} &= \frac{\partial \tilde{c}_e(x, t)}{\partial t} + \pi_1 i(t), \forall x \in \Omega_1, \\
\frac{\partial \gamma(x, t)}{\partial t} &= \frac{\partial \tilde{c}_e(x, t)}{\partial t} - \pi_3 i(t), \forall x \in \Omega_3,
\end{align*}
\]

with the dynamics in this co-ordinate system then converting (21) into the simpler, unfurcused diffusion equation

\[
\frac{\partial \gamma(x, t)}{\partial t} = D_{e,k} \frac{\partial^2 \gamma(x, t)}{\partial x^2}, \quad x \in \Omega_k, \quad k \in \{1, 3\},
\]

where the forcing reactivity term in (21) has been moved into the boundary conditions via (9d).

In each domain, the Laplace transform solution for (24) (assuming zero initial conditions) is

\[
\begin{align*}
\gamma(x, s) &= A_1(s) e^{\sqrt{\nu_1} x} + B_1(s) e^{\sqrt{\nu_2} x}, \quad x \in \Omega_1, \\
\tilde{c}_e(x, s) &= A_2(s) e^{\sqrt{\nu_1} x} + B_2(s) e^{\sqrt{\nu_2} x}, \quad x \in \Omega_2,
\end{align*}
\]

\[
\gamma(x, s) = A_3(s) e^{\sqrt{\nu_1} x} + B_3(s) e^{\sqrt{\nu_2} x}, \quad x \in \Omega_3,
\]

with spatial derivatives

\[
\begin{align*}
\frac{\partial \gamma}{\partial x} &= A_1(s) \sqrt{\nu_1} e^{\sqrt{\nu_1} x} - B_1(s) \sqrt{\nu_2} e^{\sqrt{\nu_2} x}, \quad x \in \Omega_1, \\
\frac{\partial \tilde{c}_e}{\partial x} &= A_2(s) \sqrt{\nu_1} e^{\sqrt{\nu_1} x} - B_2(s) \sqrt{\nu_2} e^{\sqrt{\nu_2} x}, \quad x \in \Omega_2, \\
\frac{\partial \gamma}{\partial x} &= A_3(s) \sqrt{\nu_1} e^{\sqrt{\nu_1} x} - B_3(s) \sqrt{\nu_2} e^{\sqrt{\nu_2} x}, \quad x \in \Omega_3.
\end{align*}
\]

The six boundary conditions for the electrolyte (9) allow the six solution coefficients \((A_1, B_1, A_2, B_2, A_3, B_3)\) of (25) to be computed.

Beginning with the no-flux condition at the current collectors (9a), then

\[
A_1(s) = B_1(s), \quad B_3(s) = A_3(s) e^{\sqrt{\nu_1} x}.
\]

The constant flux conditions (9b), (9c) at the electrode/separator means
Lastly, solution continuity at the separator interface (9d) ensures
\[ \pi_1 \frac{i(s)}{s} + A_1(s) \left( e^{\frac{s \Delta \phi_1}{v_k \epsilon}} + e^{\frac{-s \Delta \phi_1}{v_k \epsilon}} \right) = A_2(s) + B_2(s), \] (29a)

\[ A_2(s)e^{\frac{s \Delta \phi_2}{v_k \epsilon}} + B_2(s)e^{\frac{-s \Delta \phi_2}{v_k \epsilon}} = A_3(s) \left( 1 + e^{\frac{s \Delta \phi_2}{v_k \epsilon}} \right) - \pi_3 \frac{i(s)}{s}. \] (29b)

We then have four equations to solve for the four remaining unknowns \( A_1(s), A_2(s), A_3(s) \) and \( B_2(s) \).

Substituting (28a) into the continuity equation (29a) gives
\[ \frac{\pi_1 i(s)}{s} = (1 - p_1(s))A_2(s) + (1 + p_1(s))B_2(s) \] (30a)

with
\[ p_1(s) = \frac{\sqrt{D_{e,2}}}{\sqrt{D_{e,1}}} \left( e^{\frac{s \Delta \phi_1}{v_k \epsilon}} + e^{\frac{-s \Delta \phi_1}{v_k \epsilon}} \right), \] (30b)

The variables \( p_k(s), k = 1, \ldots, 7 \) are introduced to simplify the algebra. Similarly, with (28b), the continuity condition (29b) at the right hand side can be written
\[ -\pi_3 \frac{i(s)}{s} = p_2(s)A_2(s) + p_2(s)e^{\frac{s \Delta \phi_2}{v_k \epsilon}}B_2(s) \] (31a)

where
\[ p_2(s) = e^{\frac{s \Delta \phi_2}{v_k \epsilon}} \left( 1 - \sqrt{\frac{D_{e,2}}{D_{e,1}}} \left( 1 + e^{\frac{s \Delta \phi_2}{v_k \epsilon}} \right) - e^{\frac{-s \Delta \phi_2}{v_k \epsilon}} \right), \] (31b)

\[ p_3(s) = e^{\frac{-s \Delta \phi_2}{v_k \epsilon}} \left( 1 + \sqrt{\frac{D_{e,2}}{D_{e,1}}} \left( 1 + e^{\frac{s \Delta \phi_2}{v_k \epsilon}} \right) - e^{\frac{-s \Delta \phi_2}{v_k \epsilon}} \right). \] (31c)

Equating (30a) and (31a) gives an expression for \( A_2(s) \)
\[ A_2(s) = \left( \frac{\pi_1 p_3(s) + p_3(1 + p_1(s))}{s(p_2(s)(1 - p_1(s)) + 1 + p_1(s))} \right) B_2(s). \] (32a)

Subbing back into either (30a) or (31a) gives \( B_2(s) \) in terms of \( i(s) \)
\[ B_2(s) = \frac{\pi_1}{s(p_2(s)(1 - p_1(s)) + 1 + p_1(s))} \tilde{i}(s) = p_3(s) \tilde{i}(s), \] (33a)

\[ = \frac{-\pi_3}{s(p_2(s)p_3(s) + p_3(s))} \tilde{i}(s), \] (33b)

and so
\[ A_2(s) = p_4(s)p_5(s) \tilde{i}(s). \] (34)

The remaining unknowns \( A_1(s) \) and \( A_3(s) \) can then be computed from (28)
\[ A_1(s) = \frac{\sqrt{D_{e,2}}}{\sqrt{D_{e,1}}} \left( \frac{p_4(s)(p_5(s) - 1)}{e^{\frac{s \Delta \phi_1}{v_k \epsilon}} - e^{\frac{-s \Delta \phi_1}{v_k \epsilon}}} \right) \tilde{i}(s), \] (35a)

\[ A_3(s) = \frac{\sqrt{D_{e,2}}}{\sqrt{D_{e,3}}} \left( \frac{p_4(s)}{e^{\frac{s \Delta \phi_2}{v_k \epsilon}} - e^{\frac{-s \Delta \phi_2}{v_k \epsilon}}} \right) \tilde{i}(s). \] (35b)

or, more compactly,
\[ A_1(s) = p_6(s) \tilde{i}(s), \] (36a)

\[ A_3(s) = \sigma \tilde{i}(s). \] (36b)

The impedance function of the bulk electrolyte can then be expressed by
\[ \frac{v_{0c}(s)}{I(s)} = -K \left( \frac{\pi_3 + \pi_1}{s} \right) \left( e^{\frac{s \Delta \phi_1}{v_k \epsilon}} + e^{\frac{-s \Delta \phi_1}{v_k \epsilon}} \right) \tilde{i}(s), \] (37)

which is uniquely identified by the eleven parameters in Table 2. These parameters relate to the the ionic time constants, the reactivity terms and the relative diffusivities of each adjacent domain.

2.3 High-frequency response: Overpotential dynamics

The high-frequency response of the cell follows from the localised overpotential dynamics (6c) coupled to the boundary conditions (8). In \( x \in \Omega_k, k \in \{1,3\} \), the Laplace transform for this system was given in Ong and Newman (1999) as
\[ \hat{v}_k(x,s) = \frac{L_k}{v_k(s) \sinh(v_k(s))} \left( \frac{1}{\kappa} \text{cosh} \left( \frac{v_k(s)x}{L_k} \right) + \frac{1}{\kappa_s} \text{cosh} \left( v_k(s) \left( 1 - \frac{x}{L_k} \right) \right) \right), \]

with \( v_k(s) \) given by
\[ \frac{v_k^2}{L_k^2} = \frac{1}{\kappa} + \frac{1}{\kappa_s} \left( \frac{F_{0s,k} \hat{v}_{0,k}}{RT} + C_{k} \right), \] (39)

From (38), the transfer function of the overpotential dynamics, corresponding to the charge transfer resistance, can be computed as
\[ \frac{v_k(s)}{i(s)} = \left( \frac{\kappa_3/\sigma_3 + \kappa_3/\sigma_3}{\kappa_3 + \kappa_3} \right) \frac{L_1 \coth(v_3(s))}{v_3(s)} - \left( \frac{\kappa_3/\sigma_3 + \kappa_3/\sigma_3}{\kappa_3 + \kappa_3} \right) \frac{L_1 \coth(v_3(s))}{v_1(s)}. \] (40)

This function is parameterised by
\[ \theta_k^{1,k} = \frac{1}{\kappa_k + \frac{1}{\sigma_k}} \frac{F_{ax,k} \hat{v}_{0,k}}{RT}, \] (41a)

\[ \theta_k^{2,k} = \frac{A_{ck} \hat{v}_{0,k}}{\kappa_k + \sigma_k}, \] (41b)

\[ \tau_k = \frac{L_k^2}{\kappa_k + \frac{1}{\sigma_k}} C_{k}, \] (41c)
so that
\[ v_k = \theta^0_{\eta_k} + \sqrt{v_{\eta_k}} \sqrt{s} \]  
(42)
and
\[ \frac{v_\eta(s)}{I(s)} = \theta^0_{\eta_2} \frac{\coth(v_\eta(s))}{v_\eta(s)} - \theta^0_{\eta_1} \frac{\coth(v_1(s))}{v_1(s)}. \]  
(43)
Six parameters then uniquely determine the response due to heterogeneous electrode overpotentials.

2.4 Combined analysis

When combined, these results show that the current to voltage mapping of the considered decoupled and linearised battery model is characterised uniquely by the 21 parameters of Table 2. The model is then structurally identifiable with respect to these parameters and a unique solution to the parameter estimation problem for this model then exists when current-voltage data is used. The obvious next step of this research is then to develop algorithms to estimate values for these identifiable parameters from data, with the key message of this paper being that the feasibility of obtaining unique parameter estimates for these electrochemical models should be considered before any algorithms are applied.

CONCLUSIONS

A structural identifiability analysis of a decoupled and linearised Doyle-Fuller-Newman Li-ion battery electrochemical model was applied. It was shown that the model is structurally identifiable from a group of 21 parameters (composed of electrochemical quantities like the conductivities and lengths), with these parameters uniquely characterising the impedance function of this model. The parameter estimation problem for this model is then well-posed with respect to these groups of parameters. Future work will aim to exploit this result to develop an algorithm to recursively estimate the parameter values for pseudo-2D battery models from generic data.

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