Continuum-level modeling of Li-ion battery SEI by upscaling atomistically informed reaction mechanisms

Peter J. Weddle a,∗, Evan Walter Clark Spotte-Smith b,c, Ankit Verma a, Hetal D. Patel b,c, Kae Fink a, Bertrand J. Tremolet de Villers a, Maxwell C. Schulze a, Samuel M. Blau d, Kandler A. Smith a, Kristin A. Persson b,e, Andrew M. Colclasure a

a National Renewable Energy Laboratory (NREL), 15013 Denver West Parkway, Golden, CO 80401, USA
b Department of Materials Science and Engineering, University of California, Berkeley, CA 94720, USA
c Energy Technologies Area, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA
d Energy Technologies Area, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA
e Molecular Foundry, Lawrence Berkeley National Laboratory, Berkeley, CA 94720, USA

A R T I C L E   I N F O

Keywords:
Lithium-ion battery
Silicon anode
Solid-electrolyte interphase
Continuum-level model

A B S T R A C T

Understanding and controlling solid-electrolyte interphase (SEI) formation to stabilize cell performance is a significant challenge for next-generation Li-ion battery technologies. In recent years, computational modeling has become an essential tool in providing fundamental insights into SEI properties and dynamics. However, neither atomistic nor continuum-level approaches alone can capture the complexities of SEI chemistry across all relevant length and time scales. In this work, a continuum-level model is developed that is informed by reaction mechanisms obtained from first-principle calculations. The atomistically informed continuum-level model is used to understand electrolyte degradation, including the decomposition of ethylene carbonate (EC), ethyl methyl carbonate (EMC), and fluoroethylene carbonate (FEC). The model presented here is the most chemically complex continuum-level SEI model in the literature to date. The SEI model is calibrated against experimental irreversible leakage currents and shows qualitative agreement with expected SEI growth trends. The model framework is expected to accelerate fundamental understanding of SEI formation, facilitate mechanism development feedback, and dynamically interact with experimental insights.

1. Introduction

Lithium-ion batteries are the energy storage solution of choice for consumer electronics and are increasingly important for transportation (e.g., electric vehicles) [1]. For Li-ion batteries to expand use in transportation and new applications such as grid-scale energy storage, chemistries with higher energy densities are needed. However, energy-dense Li-ion batteries suffer from inherent instability. Common electrolytes based on Li salts (e.g. lithium hexafluorophosphate or LiPF6) dissolved in carbonate solvents (e.g., ethylene carbonate or EC and ethyl methyl carbonate or EMC) have limited electrochemical stability windows and are thermodynamically unstable at the plating potential of Li metal, the intercalation potentials of anodes made of graphite and silicon, and the intercalation potentials of some high-voltage cathodes [2–5].

When Li-ion electrolytes are exposed to potentials outside of their electrochemical stability windows, the electrolyte molecules degrade via parasitic side reactions. These side reactions commonly lead to the formation of nanoscale passivation films known as solid electrolyte interphase (SEI) layers [6]. Long-term Li-ion battery life is directly related to the formation of SEI on the electrochemically unstable electrode surface [2,7,8]. Ideally, the SEI, after formation, would stop growing and prevent additional Li-ion consumption and electrolyte degradation. However, in practice, SEI layers are not stagnant; SEI composition, structure, and thickness can change during cycling and calendar aging [9–12]. Continual SEI formation can result in battery failure due to loss of lithium inventory, electrolyte dry-out, and increased cell impedance [13]. Thus, an understanding of SEI formation and its resultant properties is needed to ensure long-term battery life.

In the case of graphite, SEI formation procedures and electrolyte additives have been adequately designed to such an extent that extremely long life-times are possible [3,14]. On the other hand, Si and its oxide variants suffer from continuous SEI-forming side reactions, resulting in rapid cell failure especially in terms of calendar life [15–17]. Silicon’s primary failure mechanism is commonly argued to be...
self-pulverization due to its significant volume change on cycling [18–20]. To this end, novel designs have been used to reduce the Si-particle self-pulverization [21]; for example, by reducing the Si particle size to less than a micrometer. However, cycling and calendar-life studies indicate that despite novel particle and electrode structuring to account for Si volume expansion, the SEI on Si in carbonate electrolytes is inherently non-passivating, which results in short calendar life (order 2 years) [12,15,17,22].

To increase Li-ion battery lifetimes, a primary goal of the research community is to engineer SEI layers that have (1) improved electronic passivation, (2) increased ionic conductivity, and (3) increased mechanical robustness to ensure good adhesion to the anode surface. These improvements are especially important for next-generation electrodes (e.g., Si and Li metal). Designing an optimal SEI requires knowledge of its chemically complex formation process, long-term stability, and dependencies on electrode, electrolyte composition [5,15,23,24], additives [16,25–27], operating temperature [26,27], formation rate [28], formation voltage hold [29,30], functional termination state [31,32], and formation time [3]. Such a task is extremely challenging [33], and fundamental knowledge gaps regarding the SEI remain even after decades of dedicated study.

With the aim to assist in understanding and designing better SEI layers, the present manuscript discusses a continuum-level modeling approach that communicates fundamental insights from the atomistic scale to length- and time-scales relevant to experimentally measurable signals. First, a short review of previous modeling approaches is provided to highlight the uniqueness, advantages, and disadvantages of the proposed approach. A review on experimental techniques is not provided but can be found elsewhere [3,8,10,13]. Following this review, a continuum-level single-particle model is developed to study SEI formation on a Si nanoparticle. The model includes reaction mechanisms for the decomposition of EC, EMC, and fluoroethylene carbonate (FEC) to form common SEI components (e.g., lithium fluoride (LiF), lithium carbonate or Li2CO3, and lithium ethylene dicarbonate (LED) and gas products (e.g., CO2, CH4, C2H6), with energy barriers and rate constants obtained using density functional theory (DFT), transition state theory, and Marcus theory. The model is calibrated using irreversible leakage currents from voltage-hold experiments [17,22,34–37]

The model presented here is by far the most chemically complex continuum model describing SEI formation ever reported in the literature in terms of both the number of species and the number of reactions considered. However, just as important as documenting the current model is the possibility of further developments; the model is carefully designed such that it can be modified and expanded to capture yet more complex reactivity and aid in the practical design of SEI layers.

1.1. Atomistic-level models

Atomistic modeling is well suited to identify fundamental reaction and transport mechanisms in battery environments [38], which often cannot be directly observed in experimental characterization [39]. Due to its balance of computational cost and accuracy, DFT has long been the quantum chemical method of choice for predicting reduction potentials [40], reaction thermodynamics, and kinetics [41,42]. Conventional DFT studies of electrolyte decomposition and SEI formation are conducted in a low-throughput mode [43–45], where individual reaction pathways are constructed by hand and based on human intuition. More recently, chemical reaction networks (CRNs) have been developed to automatically predict likely decomposition products [46,47] and automatically suggest chemically plausible reaction pathways to a large numbers of species [48–50], bringing a thorough exploration of battery reactivity closer to reality.

Even with advanced high-throughput and data-driven methods, DFT alone cannot capture competition between different reaction pathways or the dynamics of SEI formation and growth. Molecular dynamics (MD) can overcome this limitation to a certain extent, by predicting a time- and space-resolved picture of electrolyte reactivity. Though useful for simulating mass [51] and charge transport [52] and simulating reactivity in an unbiased manner [53], MD methods are severely limited by time scale. When classical reactive force-fields are used to generate forces and determine molecular motion, MD can readily simulate up to ≈100 ns; when ab initio methods like DFT are used, the accessible time scale is only ≈50 ps. This means that MD methods can only simulate the very early stages of SEI formation [54].

Microkinetic models using kinetic Monte Carlo (kMC) methods can bridge the gap between the atomistic and the nanoscale. These kMC models can capture both reactivity and transport, while abstracting away from an all-at-once to a (typically) molecule-scale or even more granular representation [55]. Though most kMC studies of the SEI have involved very few species and reactions and have relied on arbitrary or fitted reaction rates [56,57], it is possible to develop models with complex reaction mechanisms (for instance, based on CRN analysis) that can simulate up to the ≈1 s time scale and can reproduce and explain compositional trends in the SEI [50]. However, even with appropriate acceleration techniques, these microkinetic models struggle to access the time scales necessary to understand SEI growth, evolution, and aging (especially calendar aging), which motivates the upscaling of atomistic-level inputs to continuum-level models.

1.2. Continuum-level models

There are two phenomenological continuum-level models for the SEI. The first model, proposed by Aurbach and Zaban [58], suggests a layered SEI where inorganic decomposition products form the inner layers and organic products form the outer layers. The second model, introduced by Peled et al. [59], proposes a mosaic-type structure, where SEI species are segregated into grains and where Li-ion transport is mainly along the grain boundaries. Similar to the Aurbach-Zaban model, the Peled model assumes that the inner SEI species are more inorganic and the outer microphases are more organic. Notably both layered and mosaic structures have been experimentally observed on Li anodes [25].

Despite the general acceptance of the Aurbach-Zaban and Peled models, such layered or mosaic-like structures are not captured in most continuum-level models. Instead, simpler reaction mechanisms and SEI composition (typically of a single SEI species) are proposed [60–66]. These chemically simple SEI models follow a diffusion-limited growth dynamic where the SEI grows with a square-root-of-time dependency. The change in SEI growth rate is attributed to surface film resistance growth and associated loss of lithium inventory. Chemically simple reduced-order models have significant utility in battery lifetime models [67], but because they sacrifice specificity in the SEI composition, they are less useful in SEI engineering.

Though most continuum-level SEI models have used reduced reaction mechanisms, a small number of more chemically complex models have been proposed [9,61,68–70]. Since the present manuscript focuses on a chemically complex continuum-level model, it is worthwhile to compare and contrast these previous approaches with the present approach.

Christensen and Newman [61] and Colclasure et al. [9] implemented SEI growth models on a graphite particle. Both of these models capture EC decomposition to Li2CO3 and C2H6 gas and use a moving boundary reformulation to capture SEI growth dynamics. To capture SEI species transport, both models implement dilute-solution theory and consider mobile species (e.g., Li-ions and electrons) supported by a lattice-like SEI. In either model, the chemical complexity of the electrolyte reduction and other reactive pathways are relegated to the SEI surface sites and the bulk-phase SEI. The primary observation from the Christensen and Newman model is that the SEI growth rate depends on the electronic conductivity of the film at open-circuit conditions. Colclasure et al. [9] simulated both open-circuit voltage conditions and
In the literature to the present manuscript. The present manuscript includes orders of magnitude more homogeneous electrolyte reactions and electrolyte species than other SEI models. The additional complexity is introduced because the present manuscript does not use global reactions, but rather uses an atomistically informed reaction mechanism derived from CRN analysis and DFT calculations. The present work represents a significant jump forward in SEI modeling by providing an upscaling scheme for modeling complex reaction networks, specifically emphasizing electrolyte-phase decomposition. The present single-particle framework is most appropriate for capturing the layered structure proposed by Aurbach and Zaban [58].

2. Methodology

Fig. 1 illustrates the continuum-level model domain. Three domains are illustrated: the Si particle, the SEI, and the electrolyte. To demonstrate SEI growth dynamics, Si alloying physics are not captured [20]. The model is primarily concerned with SEI and electrolyte dynamics at a given voltage-hold (similar to Christensen and Newman's modeling at open-circuit potential [61]). As illustrated, the SEI is assumed to form a lattice structure with mobile Li-ion and electrolyte species interstitial. At the SEI/electrolyte interface, heterogeneous reactions including reduction and deposition/dissolution are simulated. In the electrolyte, species migration, diffusion, and production are simulated. In regards to production, the reaction mechanism primarily includes homogeneous reactions in the electrolyte-phase. The model internal variables are species concentrations $[X_k]$, electrostatic potential $\Phi$, and SEI thickness $\delta$.

2.1. Species conservation

Species conservation can be expressed generally for any species $k$ as

$$\frac{\partial [X_k]}{\partial t} = -\nabla \cdot J_k + \omega_k,$$

(1)

where $J_k$ is the species flux, and $\omega_k$ is the volumetric species production rate. The volumetric production rate is discussed in Section 3. The species flux $J_k$ can either be described using concentrated-solution or dilute-solution theory [72]. In dilute-solution theory, species transport results from the sum of diffusion, migration, and convection effects. The flux of species $k$ using dilute-solution theory can be expressed mathematically as

$$J_k = -D_k \nabla [X_k] - D_k \frac{z_k F}{RT} [X_k] \nabla \Phi + [X_k] \nu,$$

(2)

where $D_k$ is the species diffusion coefficient, $z_k$ is the species charge, $F$ is Faraday’s constant, $R$ is the gas constant, $\Phi$ is the potential, and $\nu$ is the bulk velocity. The bulk velocity term is important because the SEI can expand/contract significantly during a simulation, which in turn induces a bulk flow [69]. The bulk velocity due to SEI growth is handled intrinsically in the dynamic SEI-growth reference frame (Section 2.5) [73].
2.2. Electroneutrality

The local electrostatic potential $\Phi$ can either be resolved by assuming conservation of charge or by assuming electroneutrality. Conservation of charge is expressed mathematically as

$$V \cdot i + F \sum_k z_k \dot{\omega}_k = 0,$$

where $i$ is the current density. Alternatively, the electrostatic potential $\Phi$ can be solved by assuming strict electroneutrality, which can be expressed as

$$F \sum_k z_k [X_k] = 0.$$

When developing the model, both approaches in their native forms were fairly unstable and would result in either solver divergence or inability to take the first time-step. Ultimately, it was found that differentiating the strict electroneutrality equation (Eq. (5)) in time to be

$$\sum_k z_k \frac{\partial [X_k]}{\partial t} = 0,$$

resulted in the most stable governing equation to resolve potential. Importantly, “differentiating the constraint” is a viable approach if the initial species concentrations satisfy electroneutrality (Eq. (5)) [74].

2.3. SEI lattice-site conservation

For the present modeling approach, the SEI is treated as a lattice structure similar to Colclasure et al. [9]. In this lattice structure, there are three kinds of species. First, there are species that form the SEI structure (e.g., LEDC or LiF). Second, there are species that occupy the lattice sites of this SEI structure (i.e., $\text{Li}^{\text{SEI}}$). Finally, there are interstitial species that are assumed to be mobile in the SEI (i.e., $\text{Li}^{\text{SEI}}, \text{C}^{\text{SEI}}$).

In the lattice structure, the lattice site concentration is assumed to be constant [72]. Lattice-site conservation can be expressed mathematically as

$$[\text{Li}^{\text{SEI}}] + [\text{C}^{\text{SEI}}] = C,$$

where $C$ is the constant total lattice-site concentration. Differentiating this lattice-constraint equation in time to improve numerical stability results in

$$\frac{\partial [\text{Li}^{\text{SEI}}]}{\partial t} + \frac{\partial [\text{C}^{\text{SEI}}]}{\partial t} = 0.$$

Importantly, differentiating this constraint equation requires that the initial concentrations satisfy Eq. (7).

Noting that there is already a governing equation for conservation of mass (Eq. (1)), lattice conservation (Eq. (8)) adds another, which over-defines the system. Meaning, for $N$ species in the SEI there are $N + 1$ governing equations for these species. Thus, lattice-site conservation must replace a governing equation for one of the lattice-site species. The choice of which lattice species conservation equation is replaced by the lattice-site conservation is arbitrary. For the current implementation, the $\text{Li}^{\text{SEI}}$ species conservation (Eq. (1)) is replaced with the lattice-site conservation equation (Eq. (8)).

2.4. SEI growth rate

During the simulation, the SEI is able to grow/shrink due to deposition/dissolution reactions at the SEI/electrolyte interface. The SEI thickness $\delta$ is defined to spatially vary based on

$$\frac{\partial \delta}{\partial t} = \sum_{k, \text{SEI}_\text{el}} \frac{W_k}{\rho_k} \dot{s}_k,$$

where $W_k$ is the species molecular weight, $\rho_k$ is the species density, $\dot{s}_k$ is the species net surface production rate, and the summation is over SEI species that form the SEI structure (e.g., LEDC). This is the same approach as taken by Colclasure et al. [9]. This formulation assumes that the SEI thickness only changes due to interface deposition/dissolution reactions.

2.5. Moving boundary reformulation

To handle the dynamic SEI thickness changes, the model domain is transformed from a spatial mesh to a mesh that dynamically stretches/shrinks due to changes in SEI thickness. In the present model, the SEI domain is transformed from the $r - t$ domain to a $\zeta - \tau$ domain, and the electrolyte domain is transformed from the $r - t$ to a $\zeta - \tau$ domain. (Note that $\zeta$ and $\xi$ look similar, but are different symbols representing the spatially independent variable in the SEI and electrolyte domain, respectively). The new independent spatial variable $\zeta$ starts at the Si particle surface and is unity at the SEI/electrolyte interface (cf., Fig. 1). The independent variable $\zeta$ can be expressed mathematically as

$$\zeta = \frac{r - r_p}{\delta},$$

where $r$ is the independent radial variable and $r_p$ is the Si particle radius. The electrolyte independent spatial variable $\xi$ starts at the SEI/electrolyte interface and is unity at a set radial distance from the Si particle surface $R_{el}$ (cf., Fig. 1). The independent variable $\xi$ can be expressed mathematically as

$$\xi = \frac{r - r_p - \delta}{R_{el} - r_p - \delta}.$$
Note that \( R_A \) and \( r_P \) are assumed to be time invariant. From these definitions, the respective derivatives of \( \zeta \) and \( \xi \) are

\[
\left( \frac{\partial \zeta}{\partial r} \right)_t = \frac{1}{\delta_r} \cdot \left( \frac{\partial \zeta}{\partial \zeta} \right)_r = \frac{\xi}{\delta_r} \frac{\partial \zeta}{\partial \zeta}.
\]

(12)

and

\[
\left( \frac{\partial \xi}{\partial r} \right)_t = \frac{1}{R_{\xi} - r_P - \delta} \cdot \left( \frac{\partial \xi}{\partial \zeta} \right)_r = \frac{\xi - 1}{R_{\xi} - r_P - \delta} \frac{\partial \xi}{\partial \zeta}.
\]

(13)

A coordinate transformation is used to solve governing equations on the fixed mesh \([61,75,76]\). These transformation of variables can be complicated and prone to errors. So, the transformation is explained here in detail. A coordinate transformation can be expressed generally as \([75]\)

\[ A(r, t) = A \left( \zeta(r, t), \tau(t, r) \right). \]

(14)

\[
\left( \frac{\partial A}{\partial r} \right)_t = \frac{1}{\delta_r} \left( \frac{\partial A}{\partial \zeta} \right)_r + \left( \frac{\partial A}{\partial \tau} \right)_r \cdot \frac{\partial \zeta}{\partial r}.
\]

(15)

\[
\left( \frac{\partial A}{\partial r} \right)_r = \frac{\xi}{\delta_r} \frac{\partial \zeta}{\partial \tau} \left( \frac{\partial A}{\partial \zeta} \right)_r + \left( \frac{\partial A}{\partial \tau} \right) \frac{\partial \zeta}{\partial \tau}.
\]

(16)

where \( A \) is some dependent variable. For these particular transformations \( t = \tau \). The transformation of variables for the SEI domain can be expressed as

\[
\left( \frac{\partial A}{\partial r} \right)_t = \frac{1}{\delta_r} \left( \frac{\partial A}{\partial \zeta} \right)_r + \left( \frac{\partial A}{\partial \tau} \right)_r \cdot \frac{\partial \zeta}{\partial r}.
\]

(17)

\[
\left( \frac{\partial A}{\partial r} \right)_r = \frac{\xi}{\delta_r} \frac{\partial \zeta}{\partial \tau} \left( \frac{\partial A}{\partial \zeta} \right)_r + \left( \frac{\partial A}{\partial \tau} \right) \frac{\partial \zeta}{\partial \tau}.
\]

(18)

This transformation is commonly referred to as the Landau transformation \([60,73,76]\). In the electrolyte domain, the transformation of variables can be expressed as

\[
\left( \frac{\partial A}{\partial r} \right)_t = \frac{1}{R_{\xi} - r_P - \delta} \left( \frac{\partial A}{\partial \zeta} \right)_r + \left( \frac{\partial A}{\partial \tau} \right) \frac{\partial \zeta}{\partial \tau}.
\]

(19)

In the original spatial domain, species conservation can be expressed as

\[
\frac{\partial [X_k]}{\partial t} = \nabla \cdot \left( D_k \nabla [X_k] + D_k \frac{z_k F}{RT} [X_k] \nabla \Phi \right) + \dot{\omega}_k.
\]

(21)

Using the Landau transformation on SEI species conservation (Eq. (21)), the transformed SEI species conservation of species is expressed as

\[
\frac{\partial [X_k]}{\partial \tau} = -\zeta \frac{\zeta}{\delta_r} \frac{\partial [X_k]}{\partial \zeta} \frac{\partial \zeta}{\partial \tau} = \frac{1}{\delta_r} \frac{\partial [X_k]}{\partial \tau} \left( \frac{\partial [X_k]}{\partial \zeta} \right)_r + \frac{r^2_D k_z F k [X_k]}{RT} \frac{\partial \Phi}{\partial \zeta} + \dot{\omega}_k.
\]

(22)

Previous SEI model manuscripts have also used the Landau transformation to account for the moving boundary in the SEI domain \([9,60,61]\). Similarly, the transformed electrolyte species conservation can be expressed as

\[
\frac{\partial [X_k]}{\partial \tau} + \frac{\zeta - 1}{R_{\xi} - r_P - \delta} \frac{\partial [X_k]}{\partial \zeta} \frac{\partial \zeta}{\partial \tau} = \frac{1}{\delta_r} \frac{\partial [X_k]}{\partial \tau} \left( \frac{\partial [X_k]}{\partial \zeta} \right)_r + \frac{r^2_D k_z F k [X_k]}{RT} \frac{\partial \Phi}{\partial \zeta} + \dot{\omega}_k.
\]

(23)

In the fixed \( \zeta \)-transformed domain, conservation of lattice sites can be differentiated in transformed time to be

\[
\frac{\partial [\text{Li}^{0}_{\text{SEI}}]}{\partial \tau} + \frac{\partial [\text{V}^{-}_{\text{SEI}}]}{\partial \tau} = -\frac{\zeta}{\delta_r} \frac{\partial [\text{Li}^{0}_{\text{SEI}}] + [\text{V}^{-}_{\text{SEI}}]}{\partial \tau} \frac{\partial \zeta}{\partial \tau} = 0.
\]

(24)

Noting that the \( \partial \zeta/\partial \tau \) term contains lattice site conservation (i.e., the spatial derivative of a constant is zero) results in

\[
\frac{\partial [\text{Li}^{0}_{\text{SEI}}]}{\partial \tau} + \frac{\partial [\text{V}^{-}_{\text{SEI}}]}{\partial \tau} = 0.
\]

(25)

Similarly, the differentiated electroneutrality constraint (Eq. (6)) can be expressed as

\[
\sum \tau_k \frac{\partial [X_k]}{\partial \tau} = 0.
\]

(26)

Notably, Eq. (26) is the same governing equation for potential \( \Phi \) in the SEI domain using the \( \zeta - r \) transformation as it is in the electrolyte domain using the \( \zeta - \tau \) transformation. The coupled set of partial differential equations is reduced to a set of ordinary differential equations using the method of lines \([75]\), and is subsequently solved in MATLAB using the ode15s solver with a variable-dependent Mass matrix.

### 2.6. Initial conditions

The initial SEI is assumed to start as a 1 nm thick layer of LiF before the voltage-hold, which is consistent with previous continuum-level model approaches \([9,61,69]\). This initial SEI layer is required in the continuum-level model so that all phases are initiated at the beginning of the simulation. In other words, the continuum-level model does not allow for instantaneous phase creation, rather only phase growth/shrinkage. The potential equation is specified to be uniform in each domain propagating from the domain-specific boundary condition. For example, the \( \delta \) is assumed to be at a potential of \( \Phi_{\text{SEI}}(\zeta, \tau = 0) = \Phi_{\text{SEI}}(\zeta, \tau = 0) = \Phi_{\text{SEI}} \) \([60,61]\). The potential governing equation is essentially an algebraic constraint equation that implicitly depends on potential (cf., Eq. (26)), so a reasonable guess is needed to initiate the solver.

The initial species concentrations are sometimes difficult to determine in a stable manner for the SEI model. For example, if the starting concentrations of Gen2F electrolyte (1.2 M Li salt in 3:7 wt/wt EC:EMC with 3 wt.% FEC) are used as starting initial concentrations, the solver is extremely slow and may not converge to take the initial time step. This instability is mainly because uncoordinated Li-ions in solution are very unstable and quickly react to form complexes such as LiF EC, LiF EMC, and LiF FEC in solution. In other words, it is unphysical to start the SEI model with significant concentrations of uncoordinated Li-ions. However, knowing the starting concentration of coordinated complexes a priori is non-trivial. Thus, before initiating the SEI reaction model, a simple perfectly stirred reactor model is used to determine the relative concentrations of Li-coordinated complexes in the electrolyte. The perfectly stirred reactor model solves

\[
\frac{\partial [X_k]}{\partial \tau} = \dot{\omega}_k.
\]

(27)

Note that no reduction reactions are simulated in the perfectly stirred reactor model (i.e., \( \dot{\omega}_k = 0 \)). The perfectly stirred reactor model simply provides the equilibrium concentrations of Li-coordinated complexes in solution. The initial electrolyte species concentrations for the perfectly stirred reactor are specified in Table 2. Once the perfectly stirred reactor model reaches a steady-state, these steady-state concentrations are used to initialize the electrolyte species concentrations in the SEI continuum-level model. In the SEI continuum-level model, the species concentrations are assumed to be uniform spatially within each of their respective domains.

The first few time steps can be difficult for the stiff solver in the SEI continuum-level model to converge. To resolve this issue, the surface reaction net rates of progress \( \dot{\omega}_k \) and homogeneous net rates of progress \( \dot{\omega}_k \) are smoothly activated from initial condition using

\[
\Gamma = \left( 1 - 10^{-20} \right) \tanh \left( 0.05 \cdot t + 10^{-20} \right).
\]

(28)

where \( \Gamma \) modulates the net rates of progress, and \( t \) is time in seconds. This essentially "turns on" the surface reactions in a smooth manner in the first 100 s of simulation. A similar numerical trick was employed in Colclasure et al. \([9]\).
6

2.7. Boundary conditions

At the far edge of the electrolyte boundary (see right-side of Fig. 1, \( \xi = 1 \)), the electrolyte species concentration is assumed to have a no-flux condition for all non-charged species. For charged species (e.g., Li\(^+\)EC\(_{el}\), PF\(_6^-\)EC\(_{el}\), Li\(^+\)EC\(_{el}\)) at the electrolyte external boundary, the boundary condition is a constant concentration. A no-flux condition cannot be specified for charged species since a net current is developed from side reactions. Thus, a no-flux condition for charged species at the external boundary would violate conservation of current. The electrolyte potential is set to reference at the far boundary (\( \Phi_{el} |_{\xi=\infty} = 0 \)).

At the SEI/electrolyte boundary, in radial coordinates, species flux is due to both surface reactions and the moving boundary. However, in the transformed coordinate frame, the convective flux due to the moving boundary is handled separately. This means that the surface flux at the SEI/electrolyte interface is due to surface reactions only, which in \( \zeta \) space, can be expressed as

\[
\left( \mathbf{J}_i - n \left( \mathbf{X}_i \frac{\partial \delta}{\partial \zeta} \right) \right)_{\zeta=1} = - \Gamma \delta_i |_{\zeta=1},
\]

where \( \mathbf{n} \) is the unit normal vector, and \( \mathbf{J}_i \) is the species flux in the moving-boundary frame. This same boundary condition is imposed for the electrolyte domain at the \( \xi = 0 \) interface.

At the anode/SEI boundary (left-side of Fig. 1), the electron concentration is set such that the following reaction is in equilibrium

\[
e_{\text{Si}} = \ce{\text{Si}}^{\circ},
\]

where the chemical potential of the electron in the Si phase \( \mu_{e_{\text{Si}}} \) is assumed to be defined by the potential i.e.,

\[
\mu_{e_{\text{Si}}} = z_{e_{\text{Si}}} F \Phi_{\text{Si}},
\]

and the chemical potential of the electron species in the SEI \( \mu_{e_{\text{SEI}}} \) is assumed to be ideal

\[
\mu_{e_{\text{SEI}}} = \mu_{e_{\text{SEI}}}^{\circ} + RT \ln \left( \frac{e_{\text{SEI}}} {e_{\text{SEI}}^0} \right) + z_{e_{\text{SEI}}} F \Phi_{\text{SEI}},
\]

where \( e_{\text{SEI}}^{\circ} \) is the standard-state electron concentration. Assuming no potential jump between the Si and SEI, the electron concentration in the SEI at the Si interface is then

\[
\mu_{e_{\text{SEI}}} = \mu_{e_{\text{Si}}}.
\]

2.8. Species transport

The model implements dilute-solution theory to predict species transport. This is an important simplification since concentrated-solution effects are assumed to be significant in battery electrolytes [78]. However, the number of unknown diffusion parameters significantly increases if off-diagonal Onsager effects are considered (i.e., \( N \) transport properties in dilute solution theory is instead \( N \times N \) transport properties in concentrated solution theory) [72]. This is a challenge when considering a general reaction network where a significant number of electrolyte species are generated.\(^1\) Thus, a general theory is required that can approximate electrolyte species transport similar to the kinetic theory available for gas-phase species transport [72,79–82]. Species diffusion also relates to solvation shell effects and whether the ion transport is primarily due to shuttling/vehicular-transport or metal-hopping [83]. The primary transport mode in battery electrolytes has been studied using first-principles MD. These studies indicate that Li-ions in EC and DMC electrolytes can hop between solvation cages, while EC/EMC-containing electrolytes have a primarily vehicular transport mechanism [83,84]. These solvation-shell effects and associated transport also depend on the salt concentration, which can ultimately influence the measured solid-phase SEI composition [8,23,38]. At the continuum-level, these effects can be captured by simulating reversible salt-coordination/metal-hopping reactions (and associated thermodynamics) and solvated-structure transport properties. In the present

\(^1\) It is possible to reduce the number of required transport properties by only considering properties for long-lived, non-intermediate species.
study, Li-ions can coordinate to EC, EMC, and FEC, and electrolyte species diffusion is assumed to act as a dilute solution with metal-hopping reactions captured (see Table 54).

In the electrolyte, species diffusion coefficients are chosen such that the electrolyte conductivity is \( \approx 10 \text{ mS cm}^{-1} \) and the transport number is 0.4, which is reasonable for this electrolyte formulation [85,86]. To achieve these bulk phase transport properties, as a first approximation, all electrolyte species (other than PF\(_6^−\)) are assumed to have the same diffusion coefficient of \( D_{\text{Li}^+} = 9 \times 10^{-11} \text{ m}^2 \text{s}^{-1} \). The PF\(_6^−\) species is assumed to have a diffusion coefficient of 2.25E-10 m\(^2\) s\(^{-1}\), which results in a transport number of 0.4. Note that in dilute solution theory, the ionic conductivity \( \sigma \) is related to the species self-diffusion coefficients as [72]

\[
\sigma = \frac{F^2}{RT} \sum \frac{z_i^2 D_i}{[X_i]}
\]  

(35)

The simulation results are somewhat sensitive to the assumed self-diffusion coefficients. Mainly, if the diffusion coefficient is too high, then intermediate species can leave the reducing SEI surface and a build-up of intermediate products is observed in the bulk electrolyte phase. The model would be improved by implementing a more general transport theory to relax these transport assumptions [79-81].

In the SEI, the most mobile species (i.e., Li\(_{\text{SEI}}\), V\(_{\text{SEI}}\), and LiF\(_{\text{SEI}}\)) are assigned diffusion coefficients of 2E-14 m\(^2\) s\(^{-1}\), which are taken from literature [9]. The structural species (i.e., Li\(_2\)CO\(_3\))SEI, LiDF\(_{\text{SEI}}\), LiFS\(_{\text{SEI}}\), LMC\(_{\text{SEI}}\), and LEC\(_{\text{SEI}}\) (lithium ethyl carbonate) are assigned a very slow diffusion coefficient of 3E-24 m\(^2\) s\(^{-1}\). In the model, the only “free parameter” is the diffusion coefficient of the electron in the SEI phase. The electron diffusion coefficient is assumed to depend on the SEI potential and is calibrated such that the predicted leakage current is on the same scale as the measured leakage current. Since the overall SEI growth and electrolyte decomposition is governed by the electron diffusion through the SEI, the model is highly sensitive to this free parameter. All species transport properties are documented in Table 3.

3. Atomistically informed reaction mechanism

The preceding discussion has focused on formulating the SEI/electrolyte continuum-level model in general terms. That is, the model is meant to accept any given reaction mechanism developed at the atomistic scale and quickly predict SEI dynamics. The reaction mechanism – which describes the relevant phases (i.e., SEI and electrolyte), species in each phase, species thermodynamics and properties, reaction pathways, and reaction rates – is introduced to the continuum-level model via the open source CANtera package [72,87,88].

The reaction mechanism described here is obtained from atomistic analysis of elementary reaction pathways obtained via a combination of previous CRN analysis and by-hand searches. The mechanism includes pathways to the solid SEI products LDEC, Li\(_2\)CO\(_3\), LMC, LEC, and LiF, as well as the gaseous byproducts CO\(_2\), CO\(_2\)h, CH\(_4\), C\(_2\)H\(_6\), C\(_3\)H\(_6\), C\(_4\)H\(_6\), and C\(_4\)H\(_6\). Reaction thermodynamics and kinetics for homogeneous electrolyte reactions and heterogeneous reduction reactions (involving an electron from the SEI phase and a reducing species in the electrolyte phase) are obtained using DFT combined with transition state theory for (homogeneous electrolyte reactions) and Marcus theory (for reduction reactions).

While DFT and CRNs are able to provide considerable detail regarding the electrolyte phase, certain phenomena are not well captured by these methods alone. To account for the changing local solvation environments of Li\(^+\), metal hopping reactions of the type Li\(^+\) = A + B \approx A + Li – B are included, with kinetics obtained via a combination of DFT-calculated thermodynamics and Li\(^+\) residence times obtained from classical MD simulations [51,89]. More specifically, classical MD simulations have found that typical Li\(^+\) residence times are \( \approx 5 \text{ ns} \) [51], which imply an effective barrier for metal hopping of 0.266 eV. As a first approximation, all metal hopping reactions are assumed to have a 0.266 eV barrier in the exergonic direction. The endergonic rate is then computed by using the DFT-computed thermodynamics and assuming microscopic reversibility. Practically speaking, by assuming a relatively small (but not zero) barrier, the continuum-level solver is able to remain stable and the Li-ion coordinated species are near their equilibrium concentration as determined by the DFT-computed thermodynamics. Reactions for the precipitation of solid products to form the SEI and the dissolution of SEI species are assumed (see Table S2). Specifically, in the current model, five species that form in the electrolyte phase are assumed to favorably deposit to form the SEI: LDEC, Li\(_2\)CO\(_3\), LMC, LEC, and LiF.

3.1. Species thermodynamics

Thermochemical parameters for species in the electrolyte phase are calculated in the Q-Chem electronic structure code [90] using the \( \omega B97X-V \) density functional [91] and def2-TZVPDD basis set [92]. Explicit solvent shells are not considered, primarily due to the computational cost of solvent cluster calculations. Instead, solvent effects are captured implicitly using the solvent model with density (SMD) [93], with parameters appropriate for an electrolyte comprised of EC and EMC. For most species, these thermochemical parameters were previously published in the Lithium-Ion Battery Electrolyte (LIBE) dataset [94]. All species properties are assumed to follow a constant-volume equation of state for an ideal-condensed phase.

Due to the complex, heterogeneous structure and composition of the SEI, it is far more difficult to calculate the thermochemistry of species in the SEI phase. In the present model framework, the only SEI species that participate in reactions are \( \delta_{\text{SEI}} \) species and deposition products that form in the electrolyte phase. The thermodynamics of the electron species are specified such that the electrochemical reduction reactions with the electrolyte species occur nominally at the voltages reported in literature (cf., Table S2) [9,69]. The species thermodynamics of known deposition products (e.g., Li\(_{\text{SEI}}\), LDEC\(_{\text{SEI}}\), and Li\(_2\)CO\(_3\))SEI) are chosen such that the net production of deposited species is favorable. By assuming that species deposition is thermodynamically favorable, the SEI only grows. Thus, SEI “breathing” [11,95] is not currently captured in the model.

Ideally, species participating in deposition/dissolution reactions to form the SEI would have thermodynamics that are informed by measured solubility [15,96,97] or atomistic modeling [98,99]. Table S1 documents common SEI species identified in the literature with notes associated with their solubility and observed deposition locations. Of note, there can be significant differences in experimental observations on relative species solubilities (see Table S1). In principle, species solubility thermodynamics can be computed by accounting for the heat of sublimation and heat of solution to obtain the heat of dissolution [98,100]. However, these computations are highly sensitive to the local electrolyte environment and SEI surface sites, making them challenging to employ when the exact structure of the SEI and electrode surface are unknown.

3.2. Reactions

Elementary reaction pathways were obtained using the AutoTS workflow [101], which leverages the Jaguar electronic structure code [102]. Transition states were initially optimized using the \( \omega B97X-D \) density functional [103] with the def2-SVPD(-f) basis set [92] and the polarizable continuum model (PCM) implicit solvent environment [104] using the parameters for water. Each transition state was confirmed to be connected to the expected reaction endpoints by perturbing the transition state along the transition normal mode in the forwards and reverse direction and using a geometry optimization with the same \( \omega B97X-D/\text{def2-SVPD(-f)/PCM(water)} \) level of theory. For each reaction, the electronic energies of the optimized transition state and reaction endpoints were corrected with Q-Chem using
the o897X-V/def2-TZVPPD/SMD(EC/EMC) level of theory, making the calculations of energy barriers consistent with the thermochromic calculations discussed above. We note that the thermodynamics and kinetics of some reactions, particularly for EC decomposition reactions, were reported previously by Spotte-Smith, Kam, et al. [50]. Once transition-states and reaction endpoints are obtained, the reaction energy barriers are calculated as the difference in Gibbs free energy between the transition state and the reactant (for a forwards barrier) or the product (for a reverse barrier). The Eyring equation \( k \approx k_B T / h \exp(-\Delta G^*/(k_B T)) \) is then used to calculate reaction rate coefficients, where \( k_B \) is the Boltzmann constant, \( T \) is the temperature, \( \lambda \) is the reorganization energy, \( h \) is the Planck constant, and \( \Delta G \) is the reaction free energy. The free energy change of the reaction \( \Delta G \) is calculated as

\[
\Delta G = G_{\text{reduced}} - G_{\text{unreduced}} - G_e \tag{37}
\]

where \( G_{\text{reduced}} \) is the free energy of the species in the reduced state, \( G_{\text{unreduced}} \) is the free energy of the species in the unreduced state, and \( G_e \) is the free energy of the electron. The reorganization energy consists of an inner reorganization energy and an outer reorganization energy (\( \lambda = \lambda_{\text{inner}} + \lambda_{\text{outer}} \)). The inner reorganization energy \( \lambda_{\text{inner}} \) is computed using the four-point method of Nelsen [107] with DFT at the o897X-V/def2-TZVPPD/SMD(EC/EMC) level of theory. The outer reorganization energy \( \lambda_{\text{outer}} \), which depends on the bulk electrolyte environment and is as a result difficult to predict from first principles, is specified as a constant of 0.8 eV in the present study.

The total surface reaction (in m s\(^{-1}\)), can be expressed as

\[
k_{\text{surr,ET}} = \int_{R_{\text{min}}}^{R_{\text{max}}} k_{\text{ET}} \, dR \tag{38}
\]

where \( R_{\text{min}} \) is the minimum distance to the electrode (taken to be \( \approx 2 \) Å). Assuming that the second exponential does not strongly change with radius \( R \), the integral results in

\[
k_{\text{surr,ET}} \approx \frac{k_B k_B T}{\hbar \beta} \exp(-\beta R_{\text{min}}) \exp\left(-\frac{1}{4k_B T}(\lambda + \Delta G)^2 / \lambda \right). \tag{39}
\]

The electron-transfer forward rate coefficient is then

\[
k_{\text{ET}} = \frac{k_{\text{surr,ET}}}{R_{\text{min}} - R_{\text{max}}} \tag{40}
\]

The electrolyte decomposition and SEI-forming products depend significantly on the (1) equilibrium concentrations of Li-ion coordinated complexes, (2) thermodynamics of the reactants and products, and (3) reaction energy barriers. The perfectly stirred reactor (Section 2.6) is used to determine the equilibrium concentrations between Li\(^+\), LiE\(^C\), LiEMC\(^+\), and LiFEC\(^+\) in solution before introducing a reductive surface. Using the species thermodynamics (Table 3) and metal-hopping reactions (Table S4), the equilibrium concentrations before starting the SEI model are determined to be 2.95e\(^{-19}\) (Li\(^+\)), 0.630 (LiE\(^C\)), 0.544 (LiEMC\(^+\)), and 0.030 (LiFEC\(^+\)) kmol m\(^{-3}\). These concentrations can also be expressed relatively as Li-ion coordinated species versus uncoordinated species, i.e. 0.143 ([LiE\(^C\)+]/[EC]), 0.11 ([LiEMC\(^+\)+]/[EMC]), and 0.124 ([LiFEC\(^+\)+]/[FEC]).

Note that, in a real electrolyte, Li\(^+\) is typically coordinated by multiple (4–5) solvents or anions [51]. However, here it is assumed that Li\(^+\) is at any point coordinated only by one molecule to keep the number and types of species in the model tractable (i.e., so that the population of all possible solvation environments does not need to be tracked). This simplification will affect the rates of some reactions — in particular, lowering the rate of heterogeneous charge transfer reactions. However, as electrochemical reduction reactions in our model are limited by electron diffusion (see Section 4.2 below), a change in the concentration of the Li-coordinated reducing species should not significantly affect the model results.

For heterogeneous reduction reactions, the forward rate constants are determined using Marcus theory [106]. To compute the forward rate constant for electron transfer kinetics, several assumptions are required including reorganization energies, electron coupling and prefactor estimation. The electron transfer rate can be expressed as

\[
k_{\text{ET}} = \frac{k_B k_B T}{\hbar} \exp(-\beta R) \exp\left(-\frac{1}{4k_B T}(\lambda + \Delta G)^2 / \lambda \right). \tag{36}
\]

where \( k_{\text{ET}} \) is the electron transfer forward rate in s\(^{-1}\), \( k_B \) is the tunneling coefficient factor (taken as unity), \( \beta \) is the sensitivity term (taken as 1 nm\(^{-1}\)), \( k_B \) is the Boltzmann constant, \( T \) is temperature, \( \lambda \) is the reorganization energy, \( h \) is the Planck constant, and \( \Delta G \) is the reaction free energy. The free energy change of the reaction \( \Delta G \) is calculated as

\[
\Delta G = G_{\text{reduced}} - G_{\text{unreduced}} - G_e \tag{37}
\]
equilibrium potentials govern how favorably an available electron will reduce a particular nearby Li-coordinated species, as opposed to a different nearby Li-coordinated species.

Once a Li-coordinated species is reduced, several decomposition pathways are made available (see Fig. 2), which may include additional reduction of decomposed species. Competition between different decomposition routes is determined primarily by the reaction energy barriers. Fig. 2 illustrates these homogeneous barriers for each reaction as $\Delta G^\ddagger$ in eV. For example, reduced LiEMC ([S18] in Fig. 2) has four available pathways to form either LEC, LMC, or decompose to form transesterification intermediates. The kinetic barriers for these routes are 0.66 eV (LEC), 0.70 eV (LMC), or 0.44 eV and 0.48 eV to initiate transesterification. Because the barriers to form LEC and LMC are much higher than to initiate transesterification, the model predicts relatively small amounts of LEC and LMC as compared to transesterification decomposition products.

Finally, the mechanism is sensitive to the assumed solubility of intermediate/final species. In the present model, five species are selected to favorably deposit as SEI species: LEDC, Li$_2$CO$_3$, LEC, LMC, and LiF. Once a species is deposited in the SEI phase, further decomposition of this species in the electrolyte phase is prevented. For example, BC is formed from the decomposition of LMC. If LMC is assumed to favorably deposit as a SEI species, less BC is predicted to form in the electrolyte phase. The homogeneous reactions – including intramolecular, concerted, and metal-hopping reactions – are documented in the Cantera input file, Fig. 2, Table S3, and Table S4.

4. Model calibration and results

Before predicting electrolyte decomposition and resultant SEI formation, the model is calibrated with experimentally measured voltage-hold irreversible leakage currents. Specifically, the diffusion coefficient of the electron in the SEI $D_e^-$ is tuned such that the irreversible leakage current predicted by the model is on the same order of magnitude as the leakage current measured in experiments (normalized by the active material surface area). To match experimental responses, the electron diffusion coefficient through the SEI is assumed to be voltage-dependent. Once calibrated at a specified voltage, the model-predicted SEI and electrolyte species evolution is analyzed.

4.1. Voltage-hold experiment and interpretation

To calibrate the SEI model, a Si electrode (80 wt.% Paraclete Energy Silicon, 10 wt.% Timcal carbon (C45), 10 wt.% lithium polyacrylate (LiPAA) binder, $\approx 1.1$ mg/cm$^2$ total coating loading) was tested in a half-cell (14 mm Si electrode punch, 9/16'' Li foil punch, 40 $\mu$L Gen2F electrolyte, 30 °C). Triplicate cells were run under the voltage-hold protocol (described below) for 180 h at 100 mV, 175 mV, 250 mV vs. Li. The nominal Si particle size for these cells was 150 nm. The Li/Si half cells have a nominal discharge capacity of 2 mAh in the voltage window 100 mV–750 mV. Calculated volume fractions of active material Si, conductive additive C45 and binder LiPAA are 37.75%, 5.79%, and 9.16%, respectively with a porosity of 47.3%.
Table 3
Species thermodynamics and transport properties. Electrolyte thermochemical parameters are calculated in the Q-Chem [90] using the a097X-V density functional [91] and def2-TZVP/FP basis set [92]. Solvent effects are calculated implicitly using the solvent model with density (SMD) [93,94]. Electrolyte species transport properties are chosen to mimic the electrolyte conductivity and transference number.

<table>
<thead>
<tr>
<th>Species name</th>
<th>Chemical composition</th>
<th>Enthalpy, ( h^\circ/(eV) )</th>
<th>Entropy, ( s^\circ/(eV \cdot K^{-1}) )</th>
<th>Molar volume, ( V/(cm^3 \cdot mol^{-1}) )</th>
<th>Diffusion coeff., ( D/(m^2 \cdot s^{-1}) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Li_2EC</td>
<td>Li_2, E-1</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>2E-14^a</td>
</tr>
<tr>
<td>VECl</td>
<td>E-1</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>2E-14^a</td>
</tr>
<tr>
<td>Li_ClO_3EC</td>
<td>Li_2, C_1, O-3</td>
<td>-7.5920E+3</td>
<td>3.1594E-3</td>
<td>36.95</td>
<td>3E-24^a</td>
</tr>
<tr>
<td>Li_EC</td>
<td>Li_2, C_4, H-4, O_6</td>
<td>-1.7348E+5</td>
<td>4.8343E-3</td>
<td>96.20</td>
<td>3E-24^a</td>
</tr>
<tr>
<td>LiEC</td>
<td>E-1</td>
<td>-1.3090E-1</td>
<td>3.4689E-3</td>
<td>60</td>
<td>-</td>
</tr>
<tr>
<td>LIECl</td>
<td>Li_1</td>
<td>-</td>
<td>-</td>
<td>-</td>
<td>2E-14^a</td>
</tr>
<tr>
<td>LiFEC</td>
<td>Li_1, F_1</td>
<td>-3.0353E+3</td>
<td>2.0885E-3</td>
<td>5.20</td>
<td>3E-24^a</td>
</tr>
<tr>
<td>LiC_3EC</td>
<td>Li_1, C_3, H-3, O_3</td>
<td>-1.3080E+4</td>
<td>3.4072E-3</td>
<td>66.7^b</td>
<td>3E-24^a</td>
</tr>
<tr>
<td>LiC_2H_3EC</td>
<td>Li_1, C_2, H_3, O_3</td>
<td>-6.8722E+3</td>
<td>3.3421E-3</td>
<td>66.7^b</td>
<td>3E-24^a</td>
</tr>
</tbody>
</table>

(continued on next page)
accurately captures the change in magnitude and relative trends at each voltage hold. As illustrated, the model-predicted irreversible current documents the calibrated SEI electron diffusion coefficient at different holds. The SEI has a growth rate that approximates \( t \), which is observed in coin cell data between the coin cell triplicates within the 100 mV dataset showing the largest spread. The averaged data shows a higher voltage hold capacity to normalization capacity ratio at low voltage holds (high Si state-of-charge). This behavior is expected, as there is a higher propensity for SEI formation at higher states of lithiation. Initially, the current magnitudes are high and have contributions from both reversible lithiation and irreversible parasitic SEI currents [36]. Reversible current decreases continuously until the current signature is comprised mostly of parasitic current. Consequently, there is a need to deconvolve reversible lithiation from parasitic currents for accurate comparison with the present multiphase, multispecies SEI model. The general trend for the deconvoluted irreversible current indicates that the 100 mV voltage-hold has the highest irreversible current and 250 mV has the least irreversible current. Section S3 describes the experimental capacity data and deconvolution algorithm.

Fig. 3b illustrates the irreversible leakage current from the experiments and the predicted responses after \( D_{\text{el}} \) calibration. Fig. 3a illustrates the predicted SEI thickness increase during the 180 h voltage hold. The SEI has a growth rate that approximates \( t^{1/2} \) trends, which is expected for (electron) diffusion-limited growth [9,69]. Table 4 documents the calibrated SEI electron diffusion coefficient at different voltage holds. As illustrated, the model-predicted irreversible current accurately captures the change in magnitude and relative trends at each voltage hold.²

### 4.2. Solid-phase species composition

Fig. 4 illustrates the model-predicted evolution of the SEI species concentrations every 30 h during a 100 mV, 180 h voltage hold. All plots are shown as a function of normalized SEI thickness, where 0 indicates the electrode/SEI interface and unity indicates the SEI/electrolyte interface for all time (i.e., the responses are plotted w.r.t. the moving boundary variable \( \zeta \)). Eq. (10), Eq. (11), and Fig. 3a can be used to translate the normalized distance (\( \zeta \) and \( \zeta \)) to the physical distance. As illustrated, with the current mechanism, a majority of the SEI consists of LEC and LiF (see Fig. 4e and c). Minor deposits of other structural species are predicted (i.e., Li₂CO₃, LMC, LEC). The modeled mechanism currently predicts that the SEI structural species concentrations after a 100 mV voltage hold are LiF > LEC > Li₂CO₃ > LMC > LEC. The minor deposits of some structural species (e.g., Li₂CO₃, LEC, and LMC) may indicate that there are too high of barriers or unfavorable

<table>
<thead>
<tr>
<th>Table 3 (continued).</th>
</tr>
</thead>
<tbody>
<tr>
<td>58 Li₂CO₃el</td>
</tr>
<tr>
<td>59 LiF</td>
</tr>
<tr>
<td>60 CH₃</td>
</tr>
<tr>
<td>61 LiOCH₃CH₂O₃</td>
</tr>
<tr>
<td>62 LiCH₃CH₂OOC₂</td>
</tr>
<tr>
<td>63 LiCH₃OOC₂</td>
</tr>
<tr>
<td>64 LiBO₂</td>
</tr>
<tr>
<td>65 LiC₂</td>
</tr>
<tr>
<td>66 C₂H₂</td>
</tr>
<tr>
<td>67 PEC_PEO, dimer,el</td>
</tr>
<tr>
<td>68 LiFEC, lexel₁⁺</td>
</tr>
<tr>
<td>69 LiFEC, RO₂⁺</td>
</tr>
<tr>
<td>70 LiPC</td>
</tr>
<tr>
<td>71 LiPC, RO₂⁺</td>
</tr>
<tr>
<td>72 LiBC</td>
</tr>
<tr>
<td>73 LiBC, RO₂⁺</td>
</tr>
<tr>
<td>74 LiTetra₁,₁⁺</td>
</tr>
<tr>
<td>75 LiTetra,₁,₁⁺</td>
</tr>
<tr>
<td>76 PEC, dimer, closed,el</td>
</tr>
<tr>
<td>77 PEC, dimer, open, LiF,el</td>
</tr>
<tr>
<td>78 FEC, dimer,el</td>
</tr>
</tbody>
</table>

² If experimentally, an SEI thickness of >1 nm is formed during the conditioning cycles then to compare to the model, the tuned \( D_{\text{el}} \) would need to be adjusted. However, the overall trends and species production would remain the same.
Table 4
Calibrated SEI electron diffusion coefficient at different voltage holds.

<table>
<thead>
<tr>
<th>Voltage-hold</th>
<th>( D_e / (\text{m}^2 \cdot \text{s}^{-1}) )</th>
<th>SEI conductivity ( \sigma_{SEI} / (\mu\text{S m}^{-1}) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>100 mV</td>
<td>4.0 \times 10^{-18}</td>
<td>1.17</td>
</tr>
<tr>
<td>150 mV</td>
<td>1.2 \times 10^{-18}</td>
<td>1.17</td>
</tr>
<tr>
<td>175 mV</td>
<td>7.0 \times 10^{-19}</td>
<td>1.17</td>
</tr>
<tr>
<td>250 mV</td>
<td>1.0 \times 10^{-20}</td>
<td>1.17</td>
</tr>
</tbody>
</table>

\( ^a \) Conductivity is controlled by other charge species transport (i.e., \( V_{SEI}^-, \text{Li}^{1+}_{SEI} \)).

Fig. 4. Predicted SEI species concentrations during a 100 mV, 180 h voltage hold. Responses are shown at every 30 h during the hold.

between experimental results and mechanism inputs will be explored in future work. Resulting adjustments to the model framework may include, for example, introducing homogeneous reactions in the SEI phase to further decompose LEDC to form additional \( \text{Li}_2\text{CO}_3 \), \( \text{Li}_2\text{O} \), and gas-phase products.

Fig. 4a illustrates the electron concentration in the SEI. At the SEI/electrolyte interface (\( \zeta = 1 \)), the electron concentration is small. This indicates that the heterogeneous reactions are relatively fast at this interface and any available electrons are consumed. These dynamics suggest that SEI growth rate and irreversible currents are governed by electron diffusion. Importantly, operating in a diffusion-limited regime implies that model’s sensitivity to heterogeneous reduction barriers is small.

Fig. 4 shows that the model predicts an inorganic (LiF)-rich inner layer and a mixed organic (LEDC)/inorganic (LiF) outer layer. In this case, the inorganic-rich inner layer composition and thickness is strongly influenced by the assumed initial condition of a 1 nm thick SEI of LiF. Instead of assigning an inner inorganic layer through initial conditions, the present model can evolve such an inner layer by ramping the voltage from a high value to a low value. A ramped voltage profile results in favorable \( \text{LiFEC}^+ \) reduction (\( E_{eq} = 0.872 \text{ V} \)) as opposed to \( \text{LiEC}^+ \) or \( \text{LiEMC}^+ \) reduction (\( E_{eq} = 0.670 \text{ V} \) and \( E_{eq} = 0.523 \text{ V} \), respectively, cf. Table S2). Favoring the \( \text{LiFEC}^+ \) reduction pathway will result in more LiF formation, which can deposit before reducing the voltage further to then favor LEDC formation. Achieving an inorganic LiF inner layer is fairly straightforward when a voltage between 0.872 V and 0.670 V is specified. However, achieving an organic outer layer is more complex. After forming the inner layer, if the model is held
below 0.670 V, both competing reduction pathways for LiEC\(^+\) and LiFEC\(^+\) become active to form LEDC and LiF, respectively. In this case, the model predicts a mixture of organic and inorganic species in the outer layer (see Fig. 4c and e). To achieve an inorganic-rich outer layer at these low voltages, where LiFEC\(^+\) reduction is also favorable, requires one or more of the following hypothetical conditions: (1) The FEC additive is sufficiently consumed that it is no longer a significant species in the bulk electrolyte; (2) Once LEDC is deposited, the relative solubility of LEDC in solution is decreased and the solubility of LiF in solution is increased; (3) The presence of LEDC in the SEI increases the favorability of EC and EMC reduction over FEC reduction; and/or (4) The thermodynamics of solid-phase species is such that late-forming LiF diffuses to the inner inorganic layer (similar to phase segregation). Further complex boundary conditions and additional physics (see Section 5) are beyond the current scope of the manuscript, but will be considered in future work.

4.3. Electrolyte-phase species composition

Fig. 5 and Figure S1 illustrate predicted species concentrations every 30 h during a 180 h voltage hold at 100 mV. All species are plotted with respect to the normalized distance from the SEI interface out to \(R_e = 2 \mu m\) (i.e., the responses are plotted w.r.t. \(\xi\)). The model simulates 39 mobile species that evolve in space and time (see Table 3).\(^3\) Only a select few species are plotted here. Fig. 5 illustrates additional select electrolyte species concentrations. The species concentrations indicate that at the SEI/electrolyte interface (left side of Fig. 5) the initial electrolyte species are consumed and decomposition products (e.g., DMC, DEC) are produced. At early times, more current is available at the SEI interface, which results in more reduction reactions, which results in more consumption/production of electrolyte species. As the SEI thickness increases (Fig. 3a), the irreversible current decreases, which results in less overall reactivity.

Fig. 5 illustrates two predominant electrolyte dynamics. The left and middle plots show species forming at the SEI interface and diffusing away from the SEI interface. At early times, higher species concentrations are predicted; at later times when the SEI has grown and there is less irreversible current, fewer species are formed at the interface. Since these species are further consumed (see Fig. 2), they do not accumulate in the electrolyte phase. The species illustrated on the right have different dynamics than the species plotted on the left and in the center. Specifically, the species on the right are considered “final” in the mechanism, i.e., there are no reactions that consume these species favorably (note that all reactions are expressed as reversible). Because these species are not readily consumed, they accumulate in the electrolyte phase during the 180 h hold. Figure S1 illustrates additional electrolyte species formed during a 100 mV/180 h voltage hold, including gas-phase species dissolved in the electrolyte (Figure S1a-i). Under the present mechanism, all gas species are found to increase during the voltage hold.

5. Discussion

The primary goal of the present manuscript is to document a continuum-level model that can communicate atomistic knowledge (e.g., reaction pathways, rates, species transport properties) to time- and length-scales that are experimentally relevant. Additionally, the present model is meant to act as a bridge to validate (electro)chemical mechanisms proposed from experimentally observed signals. More specifically, it is common for the atomistic modeling community to propose reaction networks, decomposition rates, species transport properties, etc. that seek to explain initial Li-ion SEI formation. However, translating these predictions to long timescales has been challenging, preventing significant experimental validation. Additionally, it is common for the experimental community to detect atom/bond/molecule signals and propose (typically global) reactions that would logically produce these measured signals [109,110]. However, the mechanisms proposed from experimental observation are not typically predictive, i.e., they struggle with answering: “What would occur if additive \(X\) was included in the system?”.

5.1. Strengths of the present approach

A major strength of the present continuum-level model is its ability to operate bidirectionally, such that atomistic insights can be up-scaled to experimentally measurable signals and experimental observations can be used to refine atomistic insights. In particular, the model framework enables the use of first-principles atomistic insights to deconvolute competing reaction pathways that produce the same resultant species. Such deconvolution has traditionally challenged the experimental community. For example, vinylene carbonate (VC) is a well-documented component present in aged electrolyte whose origin is typically attributed to FEC and/or EC reduction [109,111,112]. Multiple distinct mechanistic pathways have been proposed that could feasibly result in VC evolution, and which are supported by complementary experimental analysis. Notable proposed mechanisms include the reaction of FEC with alkoxide products and the concurrent loss of HF to form VC and LiF [113,114], or alternatively the radical disproportionation of EC to form VC [115,116]. The “correct” mechanistic attribution for VC formation remains contested in the experimental literature [112], as well as consumption of VC to form additional species. CO\(_2\) presents a similar challenge, with experimentally supported pathways to produce (and consume) CO\(_2\) associated with all standard electrolyte solvents, as well as multiple downstream intermediates and – in the case of Li metal anodes – even the electrode surface itself [117].

Clearly, for such complex systems, deconvolution of all potential pathways based solely on (electro)chemical intuition is impractical. It is also unfeasible to attempt deconvolution through precise experimental detection of all complementary evolved species, when the number of relevant species is on the order of dozens to hundreds across three phases. The model, thus, offers a complementary approach to experimental measurements, whereby the favorability of competing and complex mechanistic pathways can be evaluated to both explain and help predict experimental observations.

Further, the model offers a flexible framework to capture SEI-forming reactions in all three phases. The inclusion of dissolved gas species in particular, which is likely under-explored in battery research, is unique to this model, and represents a substantial advancement in terms of realistically representing SEI behavior. The continuum model itself is somewhat mechanism-agnostic; i.e., the model may be tuned and validated on multiple different electrode and electrolyte systems and may be adjusted to incorporate varying mechanisms (as long as complete pathways and appropriate physical properties are provided).

It should be noted that the atomistic reaction network used in the present model demonstration has been developed in conjunction with experimental feedback, and as such successfully captures nuanced SEI reaction behavior that has been omitted in previous simulations. For example, the transesterification reactions and associated products presented in Fig. 2 and described in Section 3.2 have been repeatedly reported and validated by experimentalists [118–121]. Interestingly, predicting reasonable irreversible current trajectories (see Fig. 3) when transesterification reactions are present required pathways to ultimately deposit solid-phase species (see reactions labeled “FEC & Transesterification Intermediates” in Fig. 2). EMC reduction and decomposition to transesterification products produces soluble species (i.e., DMC and DEC). As a result, a significant portion of the irreversible current was taken up in bulk electrolyte changes and very little

\(^3\) Additional intermediate species are simulated with respect to only time (see Table 3).
contributed to SEI formation. If the reactions with transesterification intermediates and FEC are not included in the mechanism, then the SEI growth rate is significantly reduced (less LiF is formed) and the irreversible currents are more linear with respect to time as opposed to exhibiting a more \( t^{1/2} \) trajectory. Such pathways to “scavenge” alkoxide intermediates and form LiF have in fact been reported and validated in the experimental literature \[120\], and are captured in the present atomistic mechanism. The complexity and breadth of reaction pathways incorporated into the model is unprecedented; this facilitates improved resolution in reflecting observed (electro)chemical reactivity.

5.2. Additional SEI physics in next-generation anodes

The continuum-level model has been developed with a specific emphasis on evaluating the passivation behavior of next-generation anodes. To this end, the model has been tuned and validated on Si-based Li-ion cell chemistry. A somewhat confounding observation when
considering SEI formation and passivation is the obvious difference between SEI passivation on graphite and the lack of passivation on Si and Li surfaces. Further, while both Si and Li electrodes have significantly more volume-change on cycling than graphite, Si and Li surfaces themselves seem to have different passivation characteristics. For example, Li metal has rapid cycling fade \[8\], while nano-Si can cycle for hundreds to thousands of cycles, but has reduced calendar-life \[17\]. The following section discusses remaining challenges in understanding and simulating SEI formation in next-generation Li-ion battery materials. Some of these challenges can be addressed by the current chemically complex continuum-level model framework, while others may require incorporating additional physics. Such modifications are possible within the inherently flexible model framework, although the complexity of physics required to capture certain behaviors is high in several cases. The additional phenomena considered below are listed in rough order of increasing intensity of the adjustment(s) that would be required for the present model framework to capture.

1. **Additional reactive pathways.** As noted in Table 1, the present model and associated atomistic reaction network together represent the most chemically complex continuum-level model reported to date. However, even with the addition of orders of magnitude more chemical complexity, it is unrealistic to suggest that all possible chemical pathways will be captured. In the case of the next-generation anodes (e.g., Si and Li), important additional reactive pathways include salt decomposition; reactions of trace water; and homogeneous SEI-phase reactions (i.e., continued reactivity of species following deposition). The nature of the salt has been experimentally demonstrated to drastically influence the resulting SEI composition, particularly since salt decomposition is believed to contribute to the formation of the “pre-SEI” layer \[11\] (herein assigned via initial condition to be a 1 nm layer of LiF). Further, even in the most carefully prepared systems, the presence of trace water has been detected in both the electrolyte and as hydration water within the electrodes. Water and other protic contaminants are reported to hydrolyze electrolyte solvents at high reduction potentials \[120,122\], and in the case of Si, HF formed through reactions between water and LiPF\(_6\) salt can directly etch the Si surface \[11,122\]. Finally, the reported changes to SEI composition and morphology over repeated cycling of next-generation cells suggest that homogeneous reactions may be occurring within the SEI solid phase. Of these additional reactive pathways, incorporating salt decomposition is perhaps the most readily accessible, and this effort is already underway. Including water introduces drastically expanded mechanistic complexity and is likely to introduce numerical stability challenges, since water readily reacts with nearly all electrolyte species initially present in the system, as well as many of the reactive intermediates and products. Incorporating homogeneous SEI reactions is feasible, but would require first-principles calculations and/or experimental measurements of the thermodynamic properties and kinetic barriers to form SEI products, many of which have not yet even been conclusively identified.

2. **Capturing additional SEI dynamics.** In the present continuum-level model framework, although all reactions are fundamentally reversible, the species thermodynamics of primary deposition products has been chosen such that deposition is thermodynamically favorable, i.e., the SEI only grows. However, there is experimental evidence for dynamic SEI deposition/dissolution under conditions of both electrode cycling and storage \[11,95\]. In next-generation Si and Li materials, the electrode surface is also believed to (electro-)chemically react with the SEI, such that the SEI grows into the underlying electrode active material. Specifically for Si, cryogenic scanning transmission electron microscopy (cryo-STEM) and energy dispersive X-ray spectroscopy (EDS) elemental mapping have recently shown that the expected core–shell structure for the Si-SEI is seen after the first cycle, but after sufficient cycling, the Si-SEI becomes increasingly blended \[123\]. This behavior is in contrast to graphite, where it is believed that the SEI surface layer does not react with the underlying C\(_6\) structure and a core–shell structure is maintained \[9,61,69,124\]. The SEI’s inward growth on Si may cause additional capacity fade. In the continuum-level modeling framework, treating the first of these additional SEI dynamics (i.e., including dissolution as well as deposition) is relatively straightforward. However, this requires formulation of additional inputs including solubility values and kinetic barriers of various complex reduction products. Incorporating SEI blending with the underlying electrode would involve introducing an additional reactive boundary, which would be handled similarly to the SEI/electrolyte boundary. Instead of assuming only electrons and Li cross the electrode/SEI boundary, additional heterogeneous reactions could occur that eat away at the electrode structure and “blend” the SEI with the electrode. Adding this additional reactive boundary requires a robust understanding of the relevant heterogeneous solid–solid reactions, which is difficult to predict from first principles (see Section 1.1), and a possible reformulation to handle two moving boundaries (at the Si/SEI and SEI/electrolyte interfaces) as opposed to one (at the SEI/electrolyte interface).

3. **Chemo-mechanics.** In contrast to graphite, which undergoes relatively minor volume-change dynamics, the substantial volume change occurring in both Li and Si can influence SEI passivation via SEI cracking or thinning. However, the connection between electrode strain and SEI growth dynamics is not straightforward. In the case of Li metal, there is an “infinitely large volume expansion” \[8\], and in the case of Si, lithiation results in an expansion of \(\approx\)280% \[18\]. If these large volume changes influence SEI growth, the volume-change dependence would most likely influence cycling performance. This conclusion is supported by the reduced cycle life observed in Li-metal electrodes. For Si, efforts have been made to mitigate these volume-change effects by reducing the active particle size to the nanoscale \[125\]. Such nano-Si materials demonstrate improved cycle life but still suffer from reduced calendar life, likely due to the highly increased surface-to-volume ratios realized by the stress-reducing nano-structures. Thus, volume-change behavior alone cannot fully explain why Si exhibits reduced calendar life, since the Si expansion/contraction is relatively small during calendar-aging as compared to the expansion/contraction during cycle-aging. Including simulated finite-strain chemo-mechanics on the particle- and electrode-levels \[20,126,127\] may be necessary for the continuum-level model framework to fully capture the factors influencing SEI stability. Such addition of chemo-mechanics is not the focus of the present model – which emphasizes upscaling significantly complex reaction networks from atomistic theory – but future iterations will include finite-strain chemo-mechanics.

4. **Surface chemistry sensitivities.** One hypothesis for the differences in passivation behavior between graphite and Si or Li is that the chemical interface (Si, SiO\(_2\), or Li) exposed to the electrolyte influences the decomposition pathway. Such an argument would be supported by first-principles MD simulations that predict surface termination strongly influences initial electrolyte decomposition \[31,128\]. Surface sensitivity is supported in experimental studies where surface termination is found to influence the composition of the gas-phase species resulting from electrolyte decomposition \[129\]. Furthermore, the introduction of carbon to the silicon surface is found to affect interfacial reactivity. For example, two-dimensional silicene structures show improved
performance when covered by a thin layer of covalently bound graphene [130]. However, other measurements indicate that long-term side reactions seem to be less sensitive to electrode surface termination, and instead, are more sensitive to the initial SEI composition (e.g., whether LiF formed) [32]. Continuing the arguments against surface-termination sensitivity, Yu et al. [131] used DFT and MD to simulate EC decomposition on graphite and Li-metal and argued that EC decomposition is analogous on both surfaces, but faster for the Li surface. Logically, the surface chemistry may influence the initial electrolyte decomposition, but after an initial surface layer is formed, the underlying electrode chemistry becomes (chemically) inconsequential unless the underlying electrode/surface layer breaks apart and fresh electrode surface is exposed; for example, if Si cracks during cycling (not likely for calendar aging though). Instead, any long-term surface-mediated reactions must involve the initially formed SEI layer. Since graphite does not initially form a chemically significant different SEI than either Si or Li, it is unlikely that the surface termination of the underlying electrode has a strong effect on SEI passivation.

The reported model framework is somewhat surface-agnostic, whereby the surface is treated as an electron source and reaction favorability is driven by the availability of electrons, which is a function of potential and electron/Li diffusion through the growing SEI. At present, the model cannot readily explain or predict the differences in passivation between graphite and Si or Li anodes. Ongoing efforts are underway to determine appropriate treatment of surface chemistry, and evaluate how varying surface terminations may influence the magnitude of reaction barriers.

5. Mosaic structures. The current 1D continuum-level modeling framework captures the layered SEI structure as proposed by Arbach and Zaban [58]. However, the SEI can also form mosaic structures where SEI species are segregated into grains and where Li-ion transport is mainly regulated to the grain boundaries. To capture this mosaic structure, researchers such as Röder et al. [68] have simulated kMC instances of the SEI and proposed preferential binding energies to form the expected grain-like structures. However, additional challenges persist when formulating an SEI model that can capture the intricacies of a mosaic SEI. First, formulating a mosaic SEI model likely requires additional dimensions above a 1D framework (likely 2D or 3D) to accurately capture the SEI growth and preferential Li-ion and electron transport pathways. Second, species-specific binding energies need to be provided to the model to accurately form species-segregated grains. Third, it is likely that the (electro)chemical reactions are more sensitive to electrode surface termination, whereby the surface is treated as an electron source and reaction favorability is driven by the availability of electrons, which is a function of potential and electron/Li diffusion through the growing SEI. At present, the model cannot readily explain or predict the differences in passivation between graphite and Si or Li anodes. Ongoing efforts are underway to determine appropriate treatment of surface chemistry, and evaluate how varying surface terminations may influence the magnitude of reaction barriers.

6. Non-molecular species reactivity. Finally, recent experimental studies using isotopic labeling to evaluate gaseous decomposition products have challenged the fundamental assumption of molecular species reactivity [121,132]. Specifically, there is experimental evidence that atomic “cross-reactivity” occurs during both electrolyte-phase [121] and gas-phase [132] reactions, such that a given product species may contain, for example, carbon atoms originating from EC and hydrogen atoms originating from EMC. This novel proposed behavior has not been considered by the atomistic mechanism. To capture such physics would require substantially more complex reaction networks and appropriate isotope labeling in the Cantera input file, and is beyond the scope of the present effort – but may be an interesting avenue for future expansion.

5.3. Future model/mechanism validation

The present manuscript showcases capabilities to upscale atomistically informed mechanisms to continuum-level time- and length-scales relevant for studying the SEI. Once upscaled, the model can be validated by using experiments that detect solid-phase composition [23, 95,122,133-138], liquid-phase composition, and gas-phase composition [117,136-140]. These detection techniques can serve as a feedback mechanism to refine the atomistically informed thermodynamics, pathways, transport, and kinetics.

A key challenge for the model is the need for several (potentially unknown) properties. For example, to validate the solid-phase composition, the model requires thermodynamics and kinetics for deposition/dissolution reactions. Additionally, if there are homogeneous reactions occurring within the SEI [122,138], associated pathways, thermodynamics, and barriers are required (alongside species molar volumes). Similarly, to validate the gas-phase production, the species-specific solubility is required to communicate the species concentration in the electrolyte-phase to species concentrations in the gas-phase [141-143]. Future work will include model/mechanism validation with gas-phase detection techniques after voltage-holds alongside required species solubility measurements.

6. Summary and conclusion

A chemically complex, continuum-level Li-ion SEI model on a single anode particle is developed to understand electrolyte decomposition and solid-electrolyte interface formation on a Si nano-particle. The model considers facile decomposition for EC, EMC, and FEC to form common SEI species (e.g., LEC, Li₂CO₃, LiF) and gaseous byproducts (e.g., CO, CO₂, CH₄, C₂H₂, C₂H₄, C₂H₆, C₃H₄, C₂H₂, and C₂H₃). The model is expressed generally to accept any atomistically informed reaction mechanism. In total, the model considers 9 SEI species, 22 heterogeneous reactions at the SEI/electrolyte interface, 149 electrolyte homogeneous reactions, and 78 electrolyte species.

The model is used to predict electrolyte and SEI composition during a 180 h voltage hold. After calibration, the model correctly captures the irreversible leakage currents due to parasitic reactions at a variety of voltage-holds (e.g., 100 mV, 150 mV, 175 mV, and 250 mV). With the current mechanism, the model predicts significant LECD and LiF formation as compared to other SEI species (i.e., Li₂CO₃, LEC, LMC). Additionally, the model predicts that the SEI grows at a square-root-of-time rate due to electron diffusion-limited transport through the SEI. The model is expected to be a key tool to communicate between atomistic predictions and experiment observations.

## Nomenclature

<table>
<thead>
<tr>
<th>Variable</th>
<th>Description</th>
<th>SI Units</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>Dummy dependent variable</td>
<td>–</td>
</tr>
<tr>
<td>C</td>
<td>Total lattice-site concentration</td>
<td>kmol m⁻³</td>
</tr>
<tr>
<td>Dₖ</td>
<td>Equilibrium potential</td>
<td>V</td>
</tr>
<tr>
<td>F</td>
<td>Faraday’s constant</td>
<td>s A kmol⁻¹</td>
</tr>
<tr>
<td>ΔG</td>
<td>Reaction free energy</td>
<td>J kmol⁻¹</td>
</tr>
<tr>
<td>ΔGₛ</td>
<td>Reaction kinetic barrier</td>
<td>J kmol⁻¹</td>
</tr>
<tr>
<td>h</td>
<td>Planck’s constant</td>
<td>kg m² s⁻¹</td>
</tr>
<tr>
<td>J</td>
<td>Current density</td>
<td>A m⁻²</td>
</tr>
<tr>
<td>Jₛ</td>
<td>Species k flux</td>
<td>kmol m⁻² s⁻¹</td>
</tr>
<tr>
<td>k_B</td>
<td>Boltzmann constant</td>
<td>kg m² s⁻¹ K⁻¹</td>
</tr>
</tbody>
</table>
\[ k_f \text{ Forward reaction rate} \quad \text{kmol, m, s} \]

\[ n \text{ Surface normal} \quad - \]

\[ R \text{ Universal gas constant} \quad \text{J kmol}^{-1} \text{K}^{-1} \]

\[ R_d \text{ Electrolyte domain radius} \quad \text{m} \]

\[ R_{\text{min}} \text{ Min electrode distance for hetero. reactions} \quad \text{m} \]

\[ r \text{ Independent radial variable} \quad \text{m} \]

\[ r_p \text{ Anode particle radius} \quad \text{m} \]

\[ s_k \text{ Net surface production rate of species} \quad \text{kmol m}^{-2} \text{s}^{-1} \]

\[ t \text{ Time} \quad \text{s} \]

\[ T \text{ Temperature} \quad \text{K} \]

\[ v \text{ Bulk velocity} \quad \text{m s}^{-1} \]

\[ [X_k] \text{ Species concentration} \quad \text{kmol m}^{-3} \]

\[ W_k \text{ Molecular weight of species} \quad \text{kmol kg}^{-1} \]

\[ z_k \text{ Species} \quad \text{k} \]

\[ \beta \text{ Sensitivity term in reduction reactions} \quad \text{m}^{-1} \]

\[ f \text{ Reaction scaling variable} \quad - \]

\[ \delta \text{ SEI thickness} \quad \text{m} \]

\[ \zeta \text{ Independent radial variable in the SEI} \quad - \]

\[ \kappa_{\text{ET}} \text{ Electron transfer forward rate} \quad \text{s}^{-1} \]

\[ \kappa_{\text{ET,ET}} \text{ Total surface reaction rate} \quad \text{m s}^{-1} \]

\[ \kappa_0 \text{ Tunneling coefficient pre-factor} \quad - \]

\[ \lambda \text{ Reorganization energy} \quad \text{J kmol}^{-1} \]

\[ \lambda_{\text{inner}} \text{ Inner reorganization energy} \quad \text{J kmol}^{-1} \]

\[ \lambda_{\text{outer}} \text{ Outer reorganization energy} \quad \text{J kmol}^{-1} \]

\[ \mu_k \text{ Species k electrochemical potential} \quad \text{J kmol}^{-1} \]

\[ \mu_k^0 \text{ Species k standard-state chemical potential} \quad \text{J kmol}^{-1} \]

\[ \xi \text{ Independent radial variable in the electrolyte} \quad - \]

\[ \rho_k \text{ Species} \quad \text{k} \text{ density} \quad \text{kg m}^{-3} \]

\[ \sigma \text{ Electronic conductivity} \quad \text{S m}^{-1} \]

\[ r \text{ Transformed independent time} \quad \text{s} \]

\[ \Phi \text{ Potential} \quad \text{V} \]

\[ \omega_k \text{ Volumetric production rate of species} \quad \text{kmol m}^{-3} \text{s}^{-1} \]

CRediT authorship contribution statement

**Peter J. Weddle:** Writing, Conceptualization, Methodology, Software, Formal analysis, Investigation, Visualization. **Evan Walter Clark Spotte-Smith:** Writing, Methodology, Software, Formal analysis, Investigation, Conceptualization. **Ankit Verma:** Writing, Formal analysis, Software, Methodology. **Hetal D. Patel:** Formal analysis, Investigation, Software. **Kae Fink:** Writing, Formal analysis. **Bertrand J. Tremolet de Villers:** Writing, Formal analysis. **Maxwell C. Schulze:** Writing, Validation, Investigation. **Samuel M. Blau:** Formal analysis, Resources, Project administration, Supervision. **Kandler A. Smith:** Project administration, Resources. **Kristin A. Persson:** Conceptualization, Funding acquisition, Project administration, Supervision, Resources. **Andrew M. Colclasure:** Writing, Conceptualization, Funding acquisition, Project administration, Supervision, Resources.

Declaration of competing interest

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests: Peter Weddle reports financial support was provided by National Renewable Energy Laboratory. Evan Walter Clark Spotte-Smith reports financial support was provided by E O Lawrence Berkeley National Laboratory. Ankit Verma reports financial support was provided by National Renewable Energy Laboratory. Hetal D. Patel reports financial support was provided by E O Lawrence Berkeley National Laboratory. Kae Fink reports financial support was provided by National Renewable Energy Laboratory. Kristin A. Persson reports financial support was provided by National Renewable Energy Laboratory. Andrew M. Colclasure reports financial support was provided by National Renewable Energy Laboratory.

Data availability

Data will be made available on request.

Acknowledgments

This work is authored in part by the National Renewable Energy Laboratory, operated by Alliance for Sustainable Energy, LLC, for the U.S. Department of Energy (DOE) under Contract No. DE-AC36-08GO28308. This research was supported by the U.S. Department of Energy’s Vehicle Technologies Office under the Silicon Consortium Project, directed by Brian Cunningham, and managed by Anthony Burrell. Additional support is provided by the Kavli Energy NanoScience Institute Philomathia Graduate Student Fellowship (E.W.C.-S.) and the Laboratory Directed Research and Development Program of Lawrence Berkeley National Laboratory under U.S. Department of Energy Contract No. DE-AC02-05CH11231 (S.M.B.). Access to and assistance using the Schrödinger Suite of software tools, including Jaguar and AutoTS, was generously provided by Schrödinger, Inc. Data for this study was produced using computational resources provided by the National Energy Research Scientific Computing Center (NERSC), a U.S. Department of Energy Office of Science User Facility under Contract No. DE-AC02-05CH11231, the Eagle HPC system at the National Renewable Energy Laboratory (NREL), and the Lawrence HPC cluster at Lawrence Berkeley National Laboratory. The views expressed in the article do not necessarily represent the views of the DOE or the U.S. Government. The U.S. Government retains and the publisher, by accepting the article for publication, acknowledges that the U.S. Government retains a nonexclusive, paid-up, irrevocable, worldwide license to publish or reproduce the published form of this work, or allow others to do so, for U.S. Government purposes.

Appendix A. Supplementary data

Supplementary material related to this article can be found online at https://doi.org/10.1016/j.electacta.2023.143121.

References

[7] D. Zhao, S. Li, Regulating the performance of lithium-ion battery focus on the electrode-electrolyte interface, Front. Chem. 8 (2020) 821.
15] A. Tornheim, S.E. Trask, Z. Zhang, Evaluation of electrolyte oxidation stability on charged LiN0.05Mn0.95O2 cathode surface through potentiostatic holds, J. Electrochem. Soc. 163 (2016) A1717.
36] C.K. Chan, R. Ruffo, S.S. Hong, Y. Cui, Surface chemistry and morphology of the solid electrolyte interface on silicon nanowire lithium-ion battery anodes, J. Power Sources 189 (2009) 1132–1140.
P.J. Weddle et al.  Electrochimica Acta 468 (2023) 143121


