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A 3D Electrochemical-Thermal Coupled Model for Pouch-Type Lithium-ion Batteries with Counter-Tab Configuration

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Abstract

Lithium-ion batteries (LIBs) are the dominant energy source for electric vehicles (EVs) and battery energy storage systems (BESS) owing to their superior performance compared to other storage technologies. However, thermal management remains a critical challenge, as LIBs must operate within a tight temperature range (typically 25-40°C) and maintain spatial temperature uniformity (ideally <5°C). Emerging large-format (>50 Ah) LIB pouch cells are particularly susceptible to spatially non-uniform heat generation driven by higher current densities near their tabs, resulting in significant temperature gradients that compromise both performance and lifetime. This work presents a novel three-dimensional electrochemical-thermal (ECT) coupled model that addresses the limitations of existing ECT models in accurately predicting spatial temperature gradients across different C-rates, particularly in large-format pouch cells with counter-tab configurations that exhibit two hot spots near the cell opposite tabs. By incorporating a heat-generation distribution factor, this ECT model captures non-uniform Joule heating arising from distributed current density in the cell metal current collectors, in turn enabling accurate predictions of spatio-temporal temperatures in these large-format pouch cells. Experimental validation through voltage and temperature measurements confirms the accuracy and robustness of the proposed model, demonstrating its ability to predict the thermal characteristics of large-format pouch cells, which are increasingly adopted in EVs and BESS. The validated model is employed to conduct a comprehensive sensitivity analysis to examine how variations in electrode layer properties influence the magnitude of volumetric heat generation rate, and the role of tab geometry in controlling its spatial distribution.

Keywords: Electrochemical-thermal coupled model, non-uniform heat generation in Lithium-ion pouch cells, temperature gradients in Lithium-ion battery cells.

1. Introduction

Lithium-ion batteries (LIBs) are playing a crucial role in transportation electrification, being the preferred energy source for electric vehicles (EVs) and battery energy storage systems (BESS) due to their high energy and power densities, high-voltage capability, and low self-discharge^{1,2}. LIB cells must work within a tight temperature range (25-40°C) and maintain spatial temperature uniformity (below 5°C) across large battery packs to ensure optimal performance and lifespan. Since LIB cells generate heat during both charging and discharging processes, battery thermal management systems (BTMS) are essential for maintaining suitable thermal conditions³.

Accurate LIB temperature field prediction, especially identifying hot spots in large-format (> 50 Ah) pouch cells, is critical for achieving effective BTMS designs targeting localized heat dissipation^{4,5}. The spatial temperature distribution in large-format LIB pouch cells that are emerging in EVs is highly influenced by their spatially non-uniform heat generation and anisotropic thermophysical properties. Anisotropic heat conduction arises from the internal layered structure of LIB pouch cells, which leads to different effective thermal conductivities parallel (k_{eff}



in-plane) and perpendicular ($k_{\text{eff cross-plane}}$) to the electrode layers. Several methods have been developed to estimate these effective thermophysical properties, including our prior experimental-numerical inverse heat transfer (IHT) techniques^{6,7,8}, steady-state experiments^{9,10}, equivalent thermal circuit approaches based on the electrode layers specifications^{11,12}, and multi-length-scale thermal modeling frameworks¹³. On the other hand, an accurate estimation of the distributed heat generation across large-format pouch cells is equally essential for high-fidelity thermal modeling and BTMS design.

Equivalent circuit models (ECMs) are commonly used to describe the electrical behavior of LIBs through resistor-capacitor networks, significantly simplifying their complex electrochemical processes¹⁴. When coupled with thermal models, ECMs provide overall heat generation rate inputs to predict average cell temperatures, in some cases, with thermal feedback loops used to update the ECM parameters. Notable examples from the literature include Estevez et al.¹⁵, Tang et al.¹⁶, and Wang et al.¹⁷ for pouch cells, and Palmieri et al.¹⁸ for cylindrical cells. However, a fundamental drawback of ECM-based thermal models is their inability to capture the spatially non-uniform heat generation, particularly the localized Joule heating caused by current density concentration near the tabs of large-format pouch cells operating at high C-rates. By treating the cell as a lumped thermal system, standard ECMs cannot resolve localized hot spots that are critical for BTMS design. To enhance spatial resolution, Zhao et al.¹⁹ developed an electro-thermal ECM that accounted for thermal transport across the cell length and thickness. While this approach improved the resolution of through-thickness temperature gradients, it neglected temperature variations across the cell width. Furthermore, because ECMs have inherent fidelity limitations and cannot evaluate electrode-level electrochemical performance, determining ECM parameters requires extensive preliminary experimental testing and optimization.

High-fidelity electrochemical-thermal (ECT) coupled models primarily rely on pseudo-two-dimensional (P2D) models to resolve sub-cell internal processes, addressing the limitations of simpler ECM-based approaches. P2D models specifically account for conservation of mass and charge, resolving the electrochemical processes along a single, one-dimensional (1D) thickness direction within a sub-cell²⁰. While the standard P2D model operates in 1D, extensions to two-dimensional (2D) and three-dimensional (3D) domains are employed to capture the spatial distribution of electrochemical processes. Critically, P2D models facilitate the development of ECT coupled models by being linked with thermal models across two distinct length scales. These coupled models typically involve two domains: one dedicated to electrochemical processes (1D, 2D, or 3D at the sub-cell level), and another focused on thermal transport (1D, 2D or 3D at the cell level). Model complexity naturally scales with the dimensionality of each domain. For instance, Wang et al.²¹ coupled a 1D P2D model with a 3D thermal model for cylindrical cells, sharing averaged volumetric heat generation rate and temperature between the domains. In a similar work by Li et al.²², 3D sub-cell and cell domains shared average information of volumetric heat generation rate and temperature. An et al.²³ analyzed a 1D P2D model in conjunction with a 1D thermal model for a LIB pouch cell, examining



temperature gradients across its thickness direction. He et al.²⁴ developed a more advanced ECT model for lithium-iron-phosphate (LFP)-based pouch cells with adjacent-tab configuration, employing a plane-projection method to transfer spatial volumetric heat generation rate from a sub-cell to a cell domain, and transferring temperature information back from the cell to sub-cell domain. Lin et al.²⁵ followed a similar approach for LFP cells at C-rates as high as 5C discharge. Other studies have leveraged ECT models to refine cell-tab design and geometry. Samba et al.²⁶ developed an ECT model applied to a unique 2D sub-cell domain to evaluate tab-induced temperature gradients, without characterization of through-thickness cell thermal behavior. In a related study, Moayedi et al.²⁷ used the Newman-Tiedemann-Gu-Kim (NTGK) empirical model to assess the impact of aspect ratios on cell thermal uniformity, while disregarding the anisotropic thermal properties of the electrode stack. Optimization efforts by Zhang et al.²⁸ and Mastali et al.²⁹ have transitioned toward sub-cell to cell, 1D-2D and 1D-3D coupled models, respectively. These models provide significant computational efficiency for tab layout analysis.

While advanced ECT models – particularly those utilizing the plane-projection method – have successfully predicted cell-average temperature and minor temperature gradients for C-rates as high as 5C, their accuracy in resolving spatial temperature gradients diminishes significantly for large-format pouch cells with a counter-tab configuration operating across various C-rates. This is a critical gap, as these cells are increasingly used in EV applications and exhibit distinct temperature gradients compared to adjacent-tab layouts. While cells with adjacent-tab configurations typically manifest a single concentrated hot spot near the tabs, cells with counter-tab configuration result in the formation of two distinct hot spots at opposite ends of the cell. This bifurcation of heat concentration leads to more complex spatial temperature gradients that current models fail to capture. Therefore, there is a clear need to develop high-fidelity ECT models capable of resolving these thermal profiles in emerging large-format pouch cells across a wide range of C-rate levels.

To accurately estimate the temperature field in large-format LIB pouch cells, this study proposes an ECT coupled model that addresses the need for higher-fidelity representations that explicitly account for anisotropic thermal conductivity and detailed 3D heat-generation distributions. Unlike previous ECT coupled models that project plane-distributed volumetric heat generation rate from sub-cell to cell domains^{24,25}, the ECT model developed in this work introduces a novel heat-generation distribution factor strategy based on dominant resistive (Joule) heating processes through the cell current collectors. This strategy enables accurate prediction of 3D spatial temperature gradients in industry-relevant LIB pouch cells with counter-tab configuration operating at different levels of C-rate. Experimental validation of both voltage responses and cell surface temperatures is provided in this study.

The structure of the remainder of this manuscript is as follows. Section 2 establishes the theoretical basis by describing our ECT model, focusing on the numerical domains and governing equations. Section 3 introduces the LIB pouch cell – with nickel manganese cobalt cathode – that is the subject of this analysis. The details of the



experimental setup are provided in Section 4, followed by Section 5, which presents and compares the experimental measurements against the numerical predictions. Section 6 then focuses on a comprehensive parametric analysis to assess the impact of sub-cell electrode and tab parameters on the cell thermal response. Section 7 summarizes the key findings and contributions of the present study.

2. Three-dimensional electrochemical-thermal coupled model

This work proposes a novel three-dimensional electrochemical-thermal (ECT) coupled model, referred to hereafter as the ECT model, to concurrently predict the electrochemical and thermal dynamics of large-format pouch cells with counter-tab configuration. This ECT model employs a bi-directional coupling between 3D sub-cell and 3D cell domains via a plane-projection scheme, as depicted in Fig. 1. In addition, this model utilizes a distribution factor (df) to ensure precise spatial distribution of the volumetric heat generation rate. While the thermal model resolves conservation of energy across the 3D cell volume, the electrochemical processes are modeled using a physics-based extended P2D model applied to the 3D sub-cell (often referred to as “P4D”), capturing electrochemical variables and their temperature dependencies across the sub-cell thickness, length, and width.

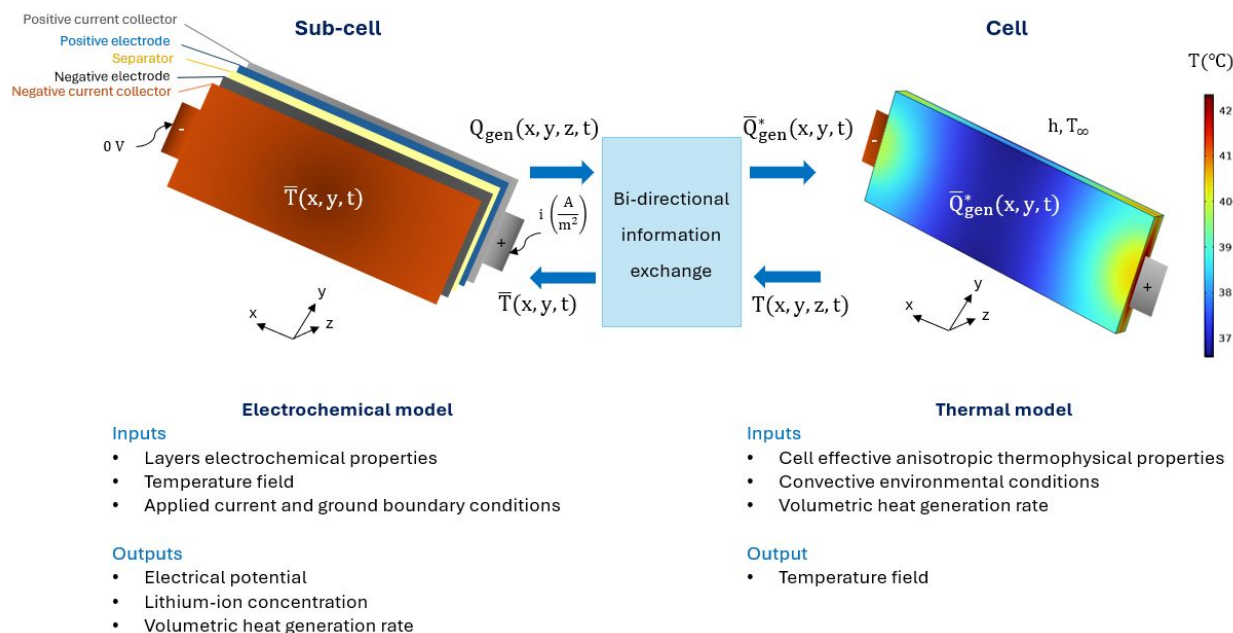


Fig. 1. Three-dimensional electrochemical-thermal (ECT) coupled model, detailing the interacting numerical domains, inputs, outputs, and bi-directional information exchange. The model utilizes a sub-cell domain for calculating electrochemical variables, such as electrical potential, lithium-ion concentration, and volumetric heat generation rate (Q_{gen}), and a cell domain for determining cell temperature field (T), where cell tabs are not considered. The coupling is bi-directional: \bar{Q}_{gen}^* is projected from the sub-cell domain to the cell domain onto the X-Y plane and a distribution factor (df), and the resulting \bar{T} is projected back to the sub-cell domain onto the X-Y plane.



This approach assumes that electrochemical processes are fully represented within a sub-cell, while the conservation of energy is analyzed over the cell domain. Volumetric heat generation rate is first calculated in the sub-cell and then projected onto the X-Y plane (sub-cell length and width) before being transferred to the cell domain. In the cell domain, temperature fields are computed and subsequently projected back to the sub-cell using a similar X-Y projection and thickness-averaged method, defined in Equation (1), to update the electrochemical processes. The mathematical expressions for these projection schemes are as follows:

$$\bar{T}(x,y,t) = \frac{1}{L_{\text{cell}}} \int_0^{L_{\text{cell}}} T(x,y,z,t) dz \quad (1)$$

$$\bar{Q}_{\text{gen}}(x,y,t) = \frac{1}{L_{\text{subcell}}} \int_0^{L_{\text{subcell}}} Q_{\text{gen}}(x,y,z,t) dz \quad (2)$$

$$\bar{Q}_{\text{gen}}^*(x,y,t) = \bar{Q}_{\text{gen}}(x,y,t) \cdot df(x,y,t) \quad (3)$$

$$df(x,y,t) = \frac{\bar{Q}_{\text{gen cc}}(x,y,t)}{Q_{\text{gen cc average}}} \quad (4)$$

$$\bar{Q}_{\text{gen cc}}(x,y,t) = \frac{1}{L_{\text{cc}}} \int_0^{L_{\text{cc}}} Q_{\text{gen cc}}(x,y,z,t) dz \quad (5)$$

Here, L_{subcell} , L_{cell} , and L_{cc} represent the thickness of the sub-cell, cell, and current collectors, respectively. As mentioned previously, the expression in Equation (2) for the projected volumetric heat generation rate is able to capture average temperatures and modest temperature gradients for cells with adjacent-tab configuration operating at high C-rates (up to 5C), as presented in ^{24,25}. However, this capability diminishes at lower C-rates and different tab layouts. This limitation arises from the formulation of Equation (2), where \bar{Q}_{gen} is computed as a thickness-averaged value. This method inherently emphasizes thicker layers and underrepresents thinner ones – particularly the current collectors, which are the thinnest layers but often the most thermally active due to concentrated current density ^{30,31}. Non-uniform current density within current collectors leads to spatial variations in Joule heating and uneven electrochemical processes in the active electrodes ³². This work's ECT model proposes a spatio-temporal distribution factor ($df(x,y,t)$) based on electric current dynamics through the current collectors, which is deemed to predominantly govern the distribution of Joule heating sources across the cell domain. This non-dimensional distribution factor is predicted by modeling the current density distribution within the metallic current collectors. The ECT model then projects this spatio-temporal distribution factor onto the volumetric heat generation rate (\bar{Q}_{gen}^*), as presented in Equation (3). This distribution factor is defined in Equations (4) and (5), considering a ratio between the projected volumetric heat generation rate in the metallic current collectors ($\bar{Q}_{\text{gen cc}}$) and the average volumetric heat generation rate ($Q_{\text{gen cc average}}$) in these layers.

2.1 Electrochemical model: P2D applied to a 3D sub-cell domain



The electrochemical model accounts for mass conservation (Equations (6) and (7) in Table 1, and Equations (19)-(21) in Table 2), charge conservation (Equations (9)-(11) in Table 1, and Equations (22)-(24) in Table 2), and electrochemical kinetics (Equations (12)-(14) in Table 1) within the sub-cell domain. This model is applied with the following assumptions:

- Active material particles in both positive and negative electrodes are considered homogeneous spheres of uniform size.
- Volume changes in the electrodes during discharge and charge processes are negligible.
- No gas is generated during the discharge and charge processes.
- All side reactions are negligible.
- Potential effects from solid electrolyte interphase (SEI) layers and lithium dendrites are negligible.

The underlying assumptions are consistent with those widely employed in similar P2D models reported in the literature^{20,33}, and are deemed appropriate for beginning-of-life (BOL) cells operating within relatively safe temperature ranges (15-50°C). Extension of this model to significantly degraded cells, which falls outside the current investigation scope, would require fundamental adjustments to account for factors such as SEI growth, side reactions, and electrodes volume expansion³⁴. A sub-cell is an architecture with porous positive and negative electrodes, separated by a porous separator, all of which are saturated with liquid electrolyte. Consequently, the movement of lithium ions (Li-ions) occurs via the liquid phase between the electrodes and the separator. At the same time, electron propagation occurs through the solid phase of the electrodes and the metallic positive and negative current collectors. The complete set of governing equations and boundary conditions is presented in Tables 1 and 2, respectively. Mass conservation is established using Fick second law of diffusion in spherical coordinates. Charge conservation is modelled using Ohm law for solid-phase electron transport and concentrated solution theory for electrolyte-based ionic transport²⁴. For these processes, the model considers the liquid volume fraction for electrolyte diffusivity and ionic conductivity (Equations 8(a) and 8(b) in Table 1), as well as the solid volume fraction for electrode ionic conductivity (Equation 8(c) in Table 1). The total volumetric heat generation rate is predicted from Equations (15)-(18) in Table 1, aggregating ohmic, activation, and reaction sources from each of the sub-cell layers, including those heat sources from positive and negative electrode layers. Note that only ohmic heating sources are considered for the metal current collector layers. While the ECT model can predict individual heat sources for each electrode, this study focuses on the total volumetric heat generation rate, which aligns with the available system-level experimental validation through cell surface temperatures. Moreover, different cell chemistries can be analyzed by simply adjusting the properties of the corresponding electrodes.



Table 1. Summary of governing equations in the electrochemical model.

Physical meaning	Governing equation	N°
Mass conservation, solid phase	$\frac{\partial c_s^\pm}{\partial t} = \frac{1}{r^2} \frac{\partial}{\partial r} \left(r^2 D_s^\pm \frac{\partial c_s^\pm}{\partial r} \right)$	(6)
Mass conservation, liquid phase	$\varepsilon_l \frac{\partial c_l^\pm}{\partial t} = \nabla \cdot (D_{l\text{eff}}^\pm \nabla c_l^\pm) + (1 - t_+) \frac{a_l i_{loc}}{F}$	(7)
Effective diffusivity and conductivity	$D_{l\text{eff}} = D_l \varepsilon_l^{1.5} \text{ (a)}; \sigma_{l\text{eff}} = \sigma_l \varepsilon_l^{1.5} \text{ (b)}; \sigma_{s\text{eff}} = \sigma_s \varepsilon_s^{1.5} \text{ (c)}$	(8)
Charge conservation, solid phase	$\nabla \cdot (\sigma_s^\pm \nabla \phi_s^\pm) = -S_a^\pm i_{loc}$	(9)
Charge conservation, liquid phase	$\nabla \cdot \left(-\sigma_{l\text{eff}}^\pm \nabla \phi_l^\pm + \frac{2RT\sigma_{l\text{eff}}^\pm}{F} \left(1 + \frac{\partial \ln f}{\partial \ln c_l^\pm} \right) (1 - t_+) \nabla \ln c_l^\pm \right) = S_a^\pm i_{loc}$	(10)
Specific area	$S_a^\pm = \frac{3\varepsilon_s^\pm}{r_p^\pm}$	(11)
Electrochemical kinetics	$i_{loc} = i_0 \left(\exp \left(\frac{\alpha_a \eta F}{RT} \right) - \exp \left(-\frac{\alpha_c \eta F}{RT} \right) \right)$	(12)
	$i_0 = F k_i c_s^{\alpha_c} (c_{s,\text{max}} - c_s)^{\alpha_a} \left(\frac{c_l}{c_{l,\text{ref}}} \right)^{\alpha_a}$	(13)
	$\eta = \phi_s - \phi_l - V_{\text{eq}}$	(14)
Volumetric heat generation rate	$Q_{\text{gen}} = Q_{\text{reaction}} + Q_{\text{activation}} + Q_{\text{ohmic}}$	(15)
	$Q_{\text{reaction}} = S_a i_{loc} T \frac{\partial E_{\text{eq}}}{\partial T}$	(16)
	$Q_{\text{activation}} = S_a i_{loc} \eta$	(17)
	$Q_{\text{ohmic}} = \sigma_s^\pm \nabla \phi_s^\pm \cdot \nabla \phi_s^\pm + \left[\sigma_{\text{eff},l}^\pm \nabla \phi_{l,k} - \frac{2\sigma_{\text{eff},l}^\pm RT}{F} \left(1 + \frac{\partial \ln f_\pm}{\partial \ln c_l^\pm} \right) (1 - t_+) \nabla (\ln c_l^\pm) \right] \cdot \nabla \phi_l^\pm$	(18)

Table 2. Summary of boundary conditions in the electrochemical model.

Physical meaning	Boundary condition	N°
Mass conservation	$-D_s^\pm \frac{\partial c_s^\pm}{\partial r} \Big _{r=0} = 0$	(19)
	$-D_s^\pm \frac{\partial c_s^\pm}{\partial r} \Big _{r=r_p} = \frac{i_{loc}}{F}$	(20)
	$-D_{l\text{eff}}^\pm \frac{\partial c_l^\pm}{\partial z} \Big _{z=L_{\text{neg cc}}} = -D_{l\text{eff}}^\pm \frac{\partial c_l^\pm}{\partial z} \Big _{z=L_{\text{subcell}}-L_{\text{pos cc}}} = 0$	(21)
Charge conservation	$-\sigma_s^\pm \cdot \nabla \phi_s^\pm \Big _{z=L_{\text{subcell}}-L_{\text{pos cc}}} = i_{\text{app}}$	(22)
	$\phi_s^\pm \Big _{z=L_{\text{neg cc}}} = 0$	(23)
	$\nabla \phi_l^\pm \Big _{z=L_{\text{neg cc}}} = \nabla \phi_l^\pm \Big _{z=L_{\text{subcell}}-L_{\text{pos cc}}} = 0$	(24)

2.2 Thermal model applied to a 3D cell domain

The thermal model is applied to the cell domain, which is treated as a homogeneous volume characterized by its effective anisotropic thermophysical properties. The model is based on the following assumptions:

- Heat generation rate and dissipation effects at the battery tabs connection are negligible.
- Thermal radiation from the battery cell surface is negligible.
- The cell effective anisotropic thermophysical properties are deemed not to change within the temperature range considered in this study.



These assumptions are deemed valid for BOL battery cells operating within relatively safe temperature ranges (15-50°C). Furthermore, the adoption of advanced metallurgical tab-welding techniques in commercial EV battery cells justifies neglecting battery tab heating effects, and these tabs are typically integrated using robust mechanisms that provide significant thermal insulation. Crucially, as EV battery packs typically operate at relatively low temperatures, convective heat flux is the dominant mode of heat dissipation, thus rendering radiation effects negligible. Therefore, the governing equation is based on the conservation of energy law and is formulated as:

$$\rho_{\text{eff}} c_{\text{peff}} \frac{\partial T}{\partial t} = \bar{Q}_{\text{gen}}^* - \nabla \cdot (-k_{\text{eff},n} \nabla T) \quad (25)$$

$$-k_{\text{eff},n} \frac{\partial T}{\partial n_S} = h(T_S - T_{\infty}) \quad (26)$$

In Equation (25), c_{peff} (J/(kgK)) denotes the cell effective specific heat capacity, ρ_{eff} (kg/m³) is the cell effective density, and $k_{\text{eff},n}$ (W/(mK)) represents the cell effective anisotropic thermal conductivities. The volumetric heat generation rate, \bar{Q}_{gen}^* (W/m³), is hierarchically transmitted from the electrochemical model. For the boundary condition represented in Equation (26), h (W/(m²K)) is the uniform heat transfer coefficient, T_S (°C) denotes the cell surface temperature, and T_{∞} (°C) represents the ambient temperature. In the present study, the ambient temperature was set to $T_{\infty}=24.5^{\circ}\text{C}$, the initial temperature to $T_{\text{initial}}=24.5^{\circ}\text{C}$, and the air conditions to 50% humidity and atmospheric pressure. The convective heat transfer coefficient (h) was set to 27 W/(m²K), a value derived from an independent air dynamics and heat transfer analysis. This coefficient represents the forced convection environment maintained within the thermal chamber during the experimental validation.

The proposed ECT model was built in COMSOL Multiphysics®, leveraging COMSOL's Lithium-ion Battery and Heat Transfer in Solids interfaces. Mesh independence analyses for the domains considered in this study, as well as specific model features, are provided in the Supplementary Information. This ECT model is novel due to its ability to accurately quantify the spatio-temporal temperature responses in pouch cells with counter-tab configuration operating at various C-rates. This capability is enabled by a derived expression for the plane-projected volumetric heat generation rate. Furthermore, this ECT model can accommodate various cell tab configurations and electrode chemistries. Adapting to different tab configurations requires modifications within the sub-cell and cell domains geometries, while incorporating different electrode chemistries necessitates updating the sub-cell layers properties to correctly estimate voltage responses and volumetric heat generation rates.

3. Pouch cell with counter-tab configuration



The cell utilized to demonstrate the ECT model proposed in Section 2 corresponds to a 63-Ah pouch-type lithium-ion battery with a nickel-manganese-cobalt (NMC) cathode, referred to hereafter as the NMC-63Ah pouch cell. A schematic of its internal layered structure is depicted in Fig. 1, on its left-hand side. Key specifications of the NMC-63Ah pouch cell are provided in Table 3. The cell internal structure consists of periodic layer arrangements (sub-cells), composed of a cathode (positive electrode), an anode (negative electrode), a separator, and positive and negative current collectors. These current collectors are made of aluminum (20 μm thick) and copper (10 μm thick), respectively. The NMC-63Ah pouch cell is formed by electrically connecting 84 of these sub-cells in a parallel configuration, with the positive and negative tabs from the current collectors exiting at opposite sides of the cell.

Table 3. Main specifications of the NMC-63Ah pouch cell. Information provided by cell manufacturer.

Feature	Value
Dimensions (mm ³)	256 x 94.5 x 13.8
Total cell area for heat dissipation (m ²)	0.058
Mass (kg)	0.895
Tab width (mm)	45
Capacity (Ah)	63 (at 1/10 C-rate)
Minimum voltage (V)	3.0
Maximum voltage (V)	4.2
Nominal voltage (V)	3.7
AC impedance at 1 kHz (m Ω)	1.5
Cathode chemistry	NMC 811
Anode chemistry	Natural graphite
Specific energy density (Wh/kg)	261
Volumetric energy density (Wh/L)	644
Cathode material ratio	97.8% active material 1.2% polyvinylidene (PVDF) binder 1.0% carbon fibers binder
Anode material ratio	95.6% active material 3.4% carboxymethyl cellulose (CMC) binder 1.0% Styrene-Butadiene Rubber (SBR) binder

The parameters used in the electrochemical model for this cell are listed in Table 4. For the separator domain, the parameters correspond to those of the electrolyte, as Li-ion transport occurs exclusively in the liquid phase within this region.

Table 4. Parameters utilized in the electrochemical model for the NMC-63Ah pouch cell.



Parameter	Cathode	Separator	Anode
Layer thickness (L, μm)	60 ^a	20 ^a	70 ^a
Particle radius (r_p , μm)	5 ^a	-	5 ^a
Solid volume fraction (ε_s)	0.45 ^a	-	0.50 ^a
Liquid volume fraction (ε_l)	0.55 ^a	0.4 ^a	0.50 ^a
Maximum Li-ion concentration ($c_{s,\text{max}}$, mol/m^3)	50,060 ^b	-	31,507 ^b
Ion migration number (t_+)	-	0.363 ^c	-
Initial Li-ion concentration (c_0 , mol/m^3)	1,000 ^b	1,000 ^b	1,000 ^b
Activation energy for diffusion (E_{ad} , J/mol)	9,978 ^d	-	68,025 ^d
Diffusion coefficient at reference temperature (D_{ref} , m^2/s)	$5.0 \cdot 10^{-13}$ ^d	-	$1.45 \cdot 10^{-13}$ ^d
Electrical/ionic conductivity (σ , S/m)	0.5 ^b	Equation (30) ^b	100 ^b
Conversion factor (a_k)	0.5 ^b	-	0.5 ^b
Transfer coefficient (α)	0.5 ^c	-	0.5 ^c
Reaction rate constant (κ , m/s)	$1 \cdot 10^{-11}$ ^e	-	$1 \cdot 10^{-11}$ ^e

^a: Information provided by cell manufacturer.

^b: Assumed from ³⁵ for a cell with the same chemistry, similar capacity, form factor, and tab configuration.

^c: Assumed from ³⁶ for a cell with the same chemistry.

^d: Assumed from COMSOL Multiphysics material database.

^e: Assumed from ³⁷ for a cell with the same chemistry.

The temperature-dependent mathematical expressions for the solid phase diffusion coefficient of the positive (D_s^+) and negative (D_s^-) electrodes, the liquid phase diffusion coefficient (D_l), and the electrolyte ionic conductivity (σ_l) are described as follows:

$$D_s^+ = D_{\text{ref}}^+ \exp\left(\frac{E_{\text{ad}}^+}{R} \cdot \left(\frac{1}{T_{\text{ref}}} - \frac{1}{T}\right)\right) \quad (27)$$

$$D_s^- = D_{\text{ref}}^- \exp\left(\frac{E_{\text{ad}}^-}{R} \cdot \left(\frac{1}{T_{\text{ref}}} - \frac{1}{T}\right)\right) \quad (28)$$

$$D_l = 10^{\left(-8.43 - \frac{54}{T-229-0.05c_1} - 2.2 \cdot 10^{-4}c_1\right)} \quad (29)$$

$$\sigma_l = 1 \cdot 10^{-4} \cdot$$

$$\left(-8.2488 + 0.053248T - 2.9871 \cdot 10^{-5}T^2 + 0.2623 \cdot 10^{-3}c_1 - 9.3063 \cdot 10^{-6}c_1T + 8.069 \cdot 10^{-9}c_1T^2 + 0.22002 \cdot 10^{-6}c_1^2\right) \quad (30)$$

Where c_l (mol/m^3) is the Li-ion concentration, F represents the Faraday constant (96,485 C/mol), η (V) quantifies the electrode overpotential, R represents the gas constant (8.314 J/(molK)), T_{ref} (298 K) is the referential temperature, and T (K) is the temperature.



This ECT model is used to simulate constant current (CC) discharge processes over the 80%-20% state-of-charge (SOC) window, and three distinct current profiles were applied to the NMC-63Ah pouch cell: 1C-rate (63 A), 1.5C-rate (94.5 A), and 2C-rate (126 A). Discharge profiles were prioritized because of their more aggressive operational conditions: the cell manufacturer specifies a maximum 2C-rate for discharge, compared to a maximum 1C-rate for charge. Evaluating high C-rates enables a critical assessment of maximum temperature and hot spots. The 80%-20% SOC window ensures the exclusion of extreme SOC conditions as per the cell manufacturer guidelines. For the thermal model, the cell effective anisotropic thermophysical properties were obtained from a multi-length-scale analysis as per our prior work¹³, and their values are $k_{\text{eff in-plane}} = 27.5 \text{ W/(mK)}$ (xx-yy direction in Fig. 1), $k_{\text{eff cross-plane}} = 0.29 \text{ W/(mK)}$ (zz direction in Fig. 1), and $c_{\text{peff}} = 1,207 \text{ J/(kgK)}$. A comparison of the experimentally measured voltage responses and transient temperatures with numerical predictions is presented in Section 5, which enables the identification of localized hot spots. Furthermore, these findings serve as a reliable basis for assessing the cell thermal behavior under different demanding conditions, including representative dynamic driving profiles³⁸.

4. Experimental setup and data acquisition

To validate the proposed ECT model, the current profiles described in the previous section were experimentally applied to the NMC-63Ah pouch cell. During testing, the cell was placed within a thermal chamber with controlled environmental conditions, and the tabs were connected to power cables to apply the current profiles. Thermocouples and voltage sensors recorded temperature and voltage transient responses, which were then compared to the numerical predictions. A brief overview of the key equipment used in the experimental setup is provided below:

- (i) Thermal chamber

The thermal chamber used in this study provides controlled ambient temperature (T_{∞}). The specific model, Espec EPX-4HF, is shown in Fig. 2(a). With an internal volume of 1,000 L, this chamber offers a temperature range from -70°C to 124°C and maintains a fluctuation below $\pm 0.5^{\circ}\text{C}$, ensuring safe and reliable conditions. The chamber is equipped with top-mounted blowers capable of $800 \text{ ft}^3/\text{min}$ airflow, enabling the maintenance of $T_{\infty}=24.5^{\circ}\text{C}$. Positioning the cell at the center of the chamber ensures nearly uniform convective conditions on both cell surfaces – an experimentally confirmed assumption that is later employed in the numerical simulations.

- (ii) Battery cell cycler

A battery cell cycler, model Chroma 17010, was utilized to apply the different current profiles to the NMC-63Ah pouch cell, enabling measurement of transient voltage and temperature responses. The cycler features 24



independent channels, each capable of delivering 0-6 V and up to 100 A. For tests requiring higher current levels, three channels were connected in parallel, allowing a current output to meet the experimental power requirements.

(iii) Temperature sensors

Thirteen T-type thermocouples (accuracy $\pm 0.5^\circ\text{C}$) were employed to measure the cell surface and ambient temperature. As shown in Fig. 2(b), 12 thermocouples were attached to the cell surfaces – six on each side – while one was placed to monitor the ambient temperature. For each test case, the cell was divided into six representative regions to capture the expected temperature gradients resulting from current concentration effects near the tabs. To ensure spatial correspondence, thermocouples T1-T6 were installed on one side, with T7-T12 placed directly opposite T1-T6, respectively, allowing symmetric thermal mapping across the cell surfaces.

(iv) Battery holder

To replicate the cell operational orientation within a real EV battery pack, a custom-made cell holder was designed and fabricated. The structure combines 80/20 aluminum framing with 3D-printed Polyethylene Terephthalate Glycol (PETG) components that support the cell along its laminated edges, ensuring vertical alignment, mechanical stability, and full exposure of the cell surfaces to the convective environment for uniform thermal conditions.

(v) Clamp connections

Electrical connections and voltage measurements were established using clamps in direct contact with the cell tabs. The power cables were secured to the clamps using standard nuts.



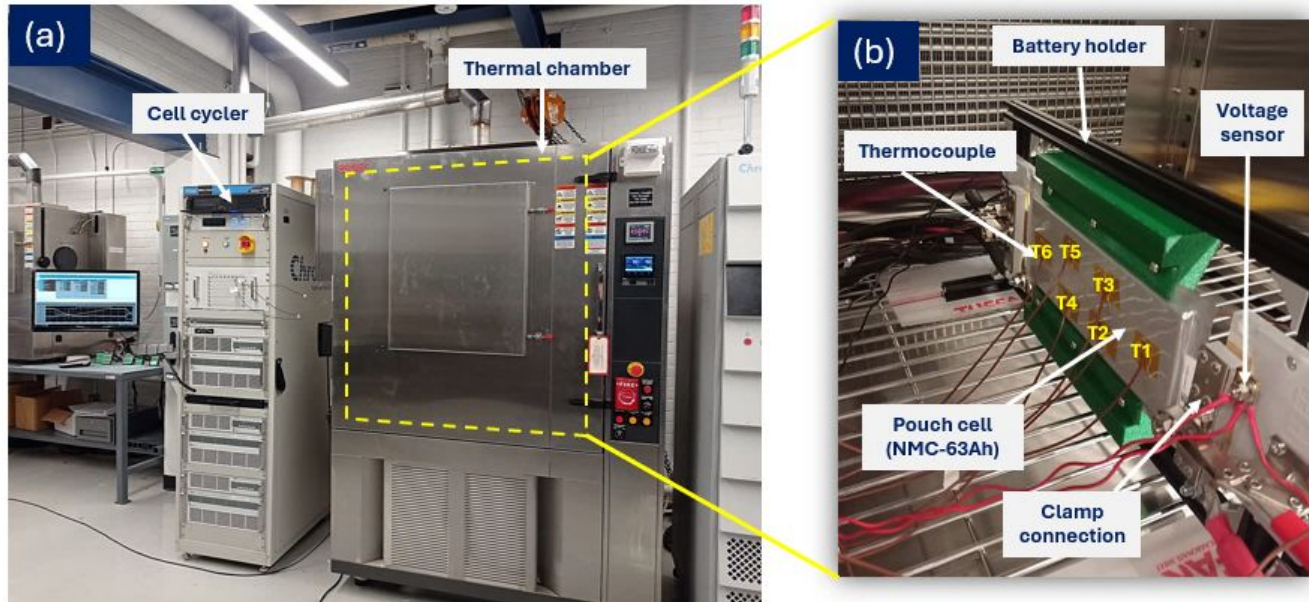


Fig. 2. Experimental setup for NMC-63Ah pouch cell testing. (a) Primary equipment, including cell cycler for current profile implementation and thermal chamber for environmental control. (b) Detailed view of vertically oriented NMC-63Ah pouch cell, illustrating placement of thermocouples on the cell surface, along with voltage sensor lead and clamp connections in battery holder.

5. Validation of the ECT model

Experimental validation of the proposed ECT model is presented in Figs. 3 and 4, which compare experimental measurements and numerical simulations for voltage responses and cell surface temperatures, respectively. To ensure repeatability of the experimental temperatures, the measurements were taken across three distinct cell samples, and error bars were included in Fig. 4 around averages. In addition, a Root Mean Square Deviation (RMSD_T) approach was implemented for the temperatures comparison to comprehensively quantify the predictive accuracy of the proposed ECT model, as defined by Equation (31) below.

$$\text{RMSD}_T = \sqrt{\frac{\sum_{t=0}^T \sum_{i=1}^N (T_{i,t,\text{exp}} - T_{i,t,\text{num}})^2}{N}} \quad (31)$$

Where N is the number of measurements (six thermocouples, every three seconds), $T_{i,t,\text{exp}}$ is the measured value for the thermocouple $\#i$ at the time t , and $T_{i,t,\text{num}}$ corresponds to the simulated value for the thermocouple $\#i$ at the time t .

Figure 3 compares the experimental voltage responses with numerical simulations from the ECT model for the C-rates considered in this study. This comparison confirms the model's high accuracy, as the numerical simulations closely match the experimental measurements. The maximum difference is 0.082V (2.1%) for the 2C-



rate scenario, occurring at the beginning of the discharge process. This may be attributed to two main factors: the deceleration of Li-ions mobility at lower temperatures³⁹, and the high polarization caused by an aggressive C-rate at the beginning and the end of discharge³⁶. The maximum difference between experimental voltage responses and numerical simulations decreases with less aggressive C-rates: 0.075V (1.9%) for 1.5C-rate, and 0.064V (1.6%) for 1C-rate. This level of accuracy in predicting voltage responses aligns well with other electrochemical-thermal models in literature, such as those from He et al.²⁴, Lin et al.²⁵, and Samba et al.²⁶, demonstrating that the proposed ECT model accurately predicts the cell electrochemical performance. This voltage prediction is essential because the magnitude and spatial distribution of the volumetric heat generation rate depend strongly on it, and it enables further analysis, such as optimizing electrode layouts.

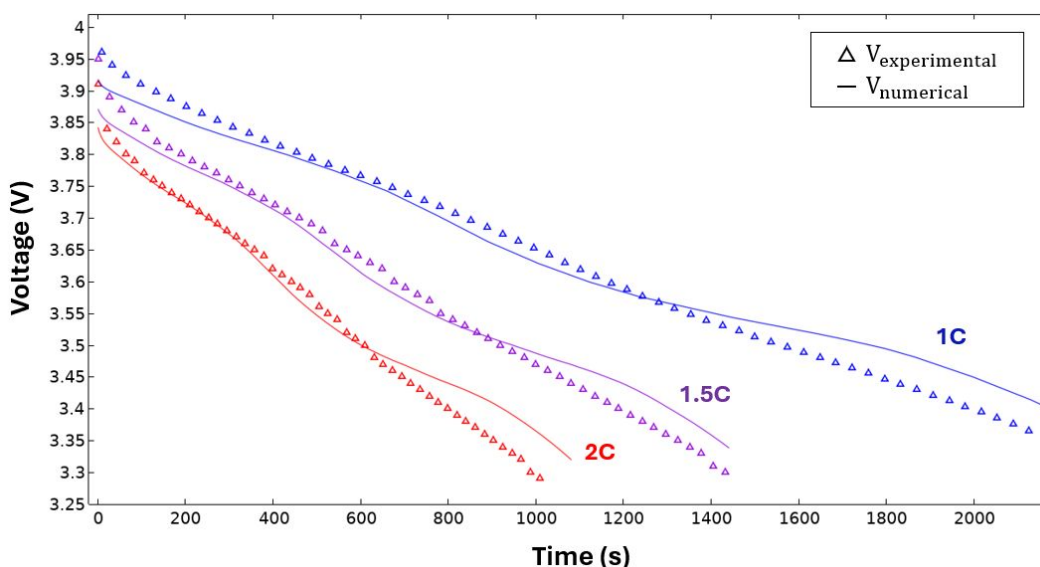


Fig. 3. Experimental measurements of voltage responses (Δ) and numerical simulations from ECT model (-) during CC discharge profiles of 1C, 1.5C, and 2C-rate applied to NMC-63Ah pouch cell across 80%-20% SOC window.

Figure 4 compares the experimental measurements of the cell surface temperatures with numerical simulations at 1C, 1.5C, and 2C rates of CC discharge. The analysis is focused on thermocouples T1 and T5, as these locations exhibit the maximum measured temperature gradients across the cells. While the experimental data were recorded every three seconds, the figure displays a subset of points to maintain visual clarity (13 points for 1C-rate, 14 points for 1.5C-rate, and 13 points for 2C-rate). The ECT model simulations show strong agreement with the measured values, remaining within the experimental error bars for most of the discharge period across all C-rates. Quantification of discrepancies via RMSD_T – calculated considering six thermocouples – yielded values of 0.404°C, 0.528°C, and 0.714°C for the 1C, 1.5C, and 2C-rate scenarios, respectively. The increase in the RMSD_T at



higher C-rates suggests that temperature prediction becomes more difficult under aggressive current profiles due to elevated volumetric heat generation rate driven by the voltage estimations. However, these discrepancies remain within acceptable limits. The maximum measured temperature at each C-rate was consistently recorded by thermocouple T1, near the positive tab, indicating localized volumetric heat generation rate driven by the concentration of current density in this region. As expected, temperature rise scales with the C-rate: the 1C-rate case exhibits an average increase of $0.15^{\circ}\text{C}/\text{min}$, while this metric is $0.77^{\circ}\text{C}/\text{min}$ for the 2C-rate scenario, 5.13 times higher for the double current profile and the same environmental convective conditions. Higher C-rates induced increasing spatial temperature gradients in the cell surface, which doubled from 0.9°C (1C-rate) to 1.8°C (2C-rate) by the end of discharge, highlighting that the NMC-63Ah pouch cell reaches non-uniform temperature fields at different current profiles for a given uniform convective environment. Overall, this validated ECT model accurately quantifies spatial temperature fields, identifying localized hot spots that contribute to non-uniform degradation and lifespan reduction. As industry trends shift toward higher C-rate capabilities, precise predictions of spatio-temporal temperatures in pouch cells are paramount, given the expected increases in maximum cell temperature and spatial temperature gradients. In particular, the development of fast-charging and high-power-discharge technologies must account for the aggressive thermal conditions experienced by large-format pouch cells.

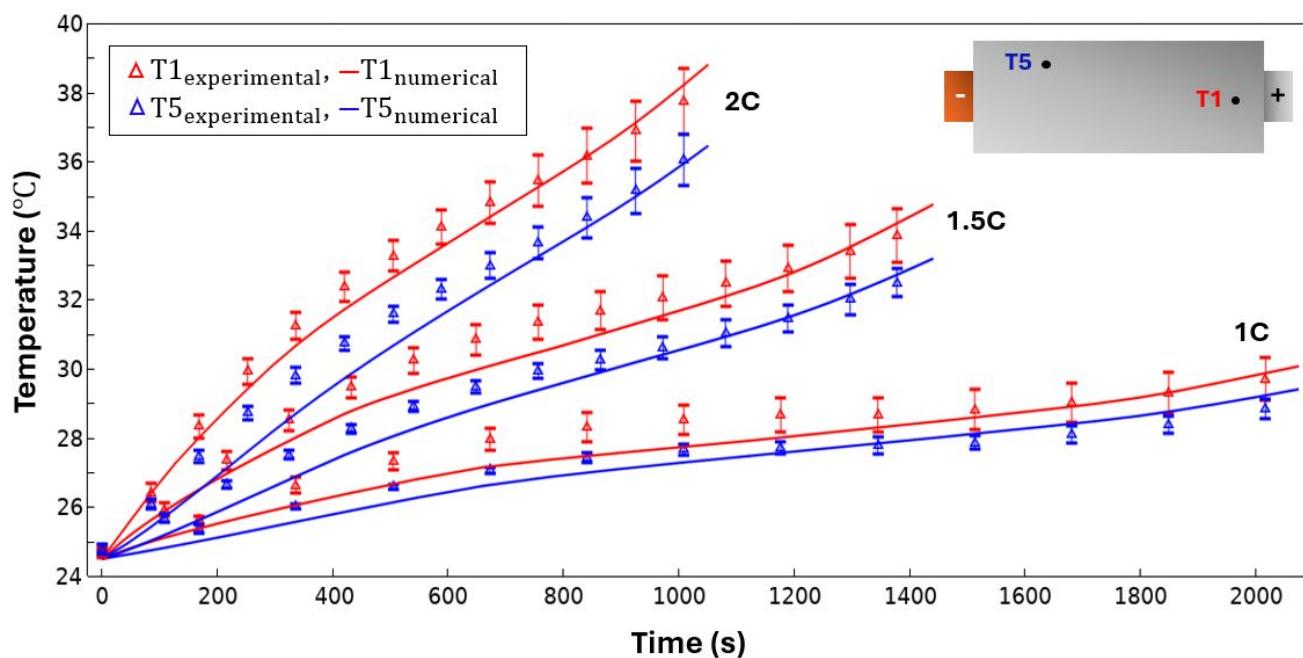


Fig. 4. Transient temperature measurements (Δ) and numerical simulations from ECT model (-) for thermocouples T1 and T5 during CC discharge profiles of 1C, 1.5C, and 2C-rate applied to NMC-63Ah pouch cell across 80%-20% SOC window. Experimental measurements represent the average of three different cells with standard deviation shown as error bars.



Figure 5 illustrates three-dimensional (3D) contours of volumetric heat generation rates (Q_{gen} , in kW/m^3) and temperature fields (T , in $^{\circ}\text{C}$) obtained from the ECT model at the end of discharge (SOC 20%) for 1C-rate, 1.5C-rate, and 2C-rate. As is presented in Figs. 5(a) to (c), the volumetric heat generation rate predominantly concentrates near the cell tabs. Specifically, the region adjacent to the positive tab exhibits the maximum Q_{gen} across the tests, which is consistent with the higher temperature measurements of thermocouple T1 (Fig. 4). The localized volumetric heat generation rate is driven by the current density concentration. At the 1C-rate scenario, the highest current density near the positive tab is 1.57 MA/m^2 , causing a maximum Q_{gen} of 70 kW/m^3 , whereas the highest current density in this region is 3.14 MA/m^2 for the 2C-rate, yielding a maximum Q_{gen} of 280 kW/m^3 , results that are aligned with the quadratic relation between current profile and magnitude of volumetric heat generation rate. In addition to the Q_{gen} intensity, the level of non-uniformity is also exacerbated with higher C-rates. For instance, the maximum Q_{gen} of 70 kW/m^3 for the 1C-rate case corresponds to 3.9 times the average (18 kW/m^3), whereas the maximum Q_{gen} of 280 kW/m^3 in the 2C-rate case is 4.5 times the average (62 kW/m^3). The non-uniformity of Q_{gen} triggers non-uniform temperature fields throughout the cell domain. As Fig. 5(d) presents for the 1C-rate scenario, the maximum temperature (T_{max}) remains within the safety limits with a value of 31.3°C , having spatial uniformity with a maximum spatial temperature gradient (ΔT_{max}) of 1.7°C . On the other hand, the optimal thermal conditions are exceeded at the 2C-rate case. Whereas T_{max} reaches 42.4°C , ΔT_{max} is 5.9°C , conditions that are detrimental to cell safety and reliability, as these levels for T_{max} and ΔT_{max} are known to accelerate non-uniform cell degradation. These 3D analyses of Q_{gen} and temperature fields in the NMC-63Ah pouch cell highlight the necessity of (i) reduction of current density from the cell layers layout – a topic that is analyzed in the next section – and (ii) BTMS strategies aimed at localized heat dissipation. Further investigation of (i) and (ii) can be directly guided by this ECT model thanks to its ability to predict spatio-temporal temperature fields and localized hot spots, enabling extensions of thermal modeling studies to module and pack levels where the through-thickness thermal gradients become more relevant due to the inter-cell materials, such as foam layers and metal fins. Furthermore, the anticipated progression toward higher C-rate capabilities driven by fast-charging/discharging requirements will impose more severe thermal stress on pouch cells, underscoring the need for further investigation into the above-mentioned topics (i) and (ii).



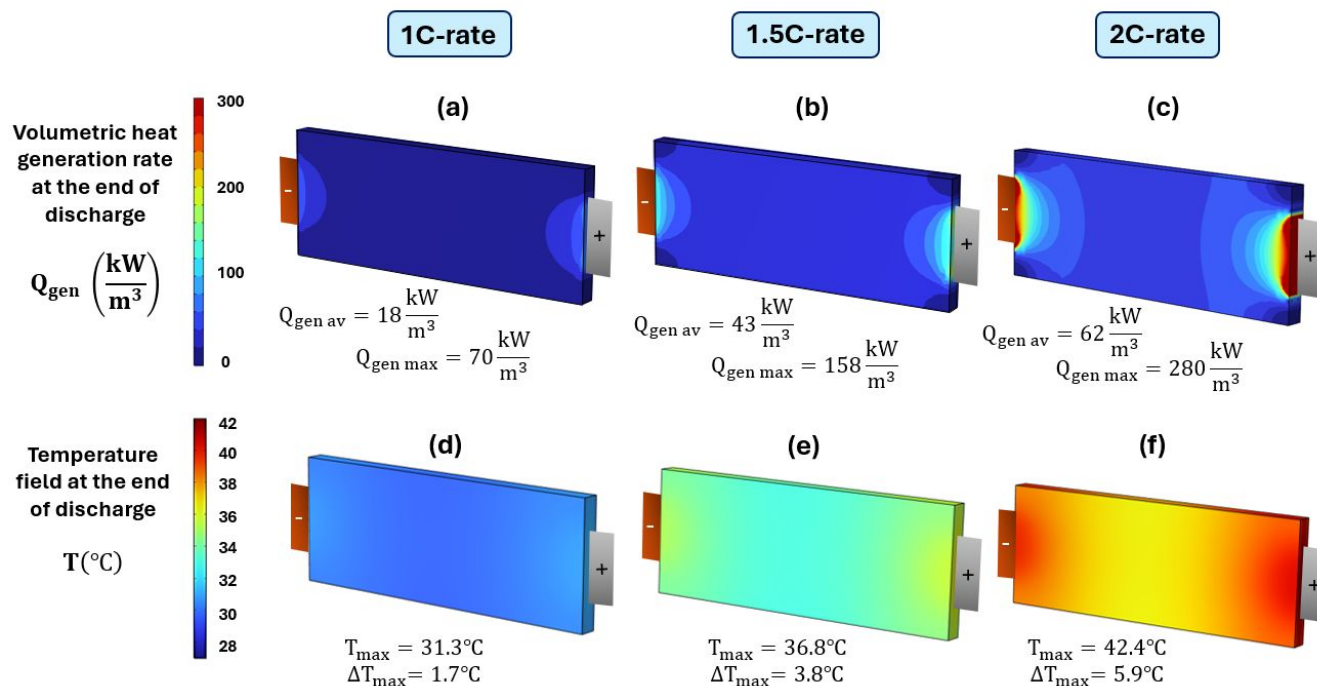


Fig. 5. Three-dimensional volumetric heat generation rates (Q_{gen}) and temperature fields at end of discharge (SOC 20%) for NMC-63Ah pouch cell under different CC discharge profiles. Contours of Q_{gen} are presented in (a) 1C-rate, (b) 1.5C-rate, and (c) 2C-rate, detailing average ($Q_{gen\ av}$) and maximum values ($Q_{gen\ max}$). Temperature fields are illustrated in (d) 1C-rate, (e) 1.5C-rate, and (f) 2C-rate, indicating maximum temperature (T_{max}) and maximum spatial temperature gradient (ΔT_{max}).

6. Sensitivity of Cell Thermal Performance to Select Electrode Parameters and Cell Tabs

To assess how modifications at the sub-cell layers affect the overall cell thermal performance, this section presents a sensitivity analysis focused on the electrode layers and cell tabs. The electrode parameters listed in Table 4 and tab width presented in Table 3 are taken as the baseline, with variations applied one at a time to isolate their individual effects. The parameters selected for variation include electrode thickness and solid volume fraction due to their influence on volumetric heat generation rate magnitude (Equations (15)-(18) in Table 1), and cell tab width as per their impact on the current density concentration. The effects of electrode particle radius (which affects the effective specific surface area in the numerical model) are discussed in the Supplementary Information (Fig. A2). This sensitivity analysis assumes the same environmental convective conditions from the experimental setup ($h = 27 \text{ W}/(\text{m}^2\text{K})$ and $T_{\infty} = 24.5^{\circ}\text{C}$), and evaluates the cell volumetric heat generation rate (Q_{gen}), maximum temperature (T_{max}), and maximum spatial temperature gradient (ΔT_{max}) during a profile of CC discharge at 2C-rate (the most aggressive current) across the SOC 80%-20% window. Also, although their variations are relatively small, the cell effective thermophysical properties were updated with the electrode thickness and solid volume fraction, as



presented in our previous article ¹³. The specific parameter values used in this sensitivity analysis are provided in Table 5:

Table 5. Selected parameters for the sensitivity analysis. The lower and upper electrode values around the baseline scenario (Table 4) cover typical ranges for these parameters reported in the literature ^{40,41}.

Parameter	Values for evaluation
Cathode thickness, L_{pos} (μm)	35, 60 ^a , 80
Anode thickness, L_{neg} (μm)	42, 70 ^a , 102
Cathode solid volume fraction, ϵ_s^+	0.35, 0.45 ^a , 0.75
Anode solid volume fraction, ϵ_s^-	0.40, 0.50 ^a , 0.68
Cell tab width (mm)	30, 45 ^a , 70

^a: Baseline value

6.1 Effect of Electrode Thickness

Figure 6 presents how variations in the thickness of cathode (L_{pos}) and anode (L_{neg}) layers impact on the volumetric heat generation rate, Q_{gen} , and thermal response, T_{max} and ΔT_{max} . The results indicate that Q_{gen} is more sensitive to the cathode layer thickness (L_{pos} , in Fig. 6(a-1)) than the anode layer thickness (L_{neg} , in Fig. 6(b-1)) during the discharge process. This finding aligns with our previous work of thermal transport across multiple length scales in EVs ¹³. The main reason for this behavior is as follows: while increasing the cathode thickness directly increases the cell capacity, it concurrently affects two factors that elevate the volumetric heat generation rate. First, it extends the Li-ion propagation path during discharge, which increases the cell internal resistance. Second, a thicker cathode layer means that more available Li-ions come from the anode in a discharge process, which further contributes to elevated volumetric heat generation rate. In turn, this greater volumetric heat generation rate drives higher values for both T_{max} and ΔT_{max} . This effect is demonstrated in Figs. 6(a-2) and 6(a-3), where a cathode layer thickness of 80 μm results in thermal conditions that are detrimental to cell lifespan, performance, and safety. From a thermal-design perspective, these results indicate that thinner cathode layers are thermally preferable when targeting a fixed capacity, as they mitigate excessive volumetric heat generation rate and temperature non-uniformity.



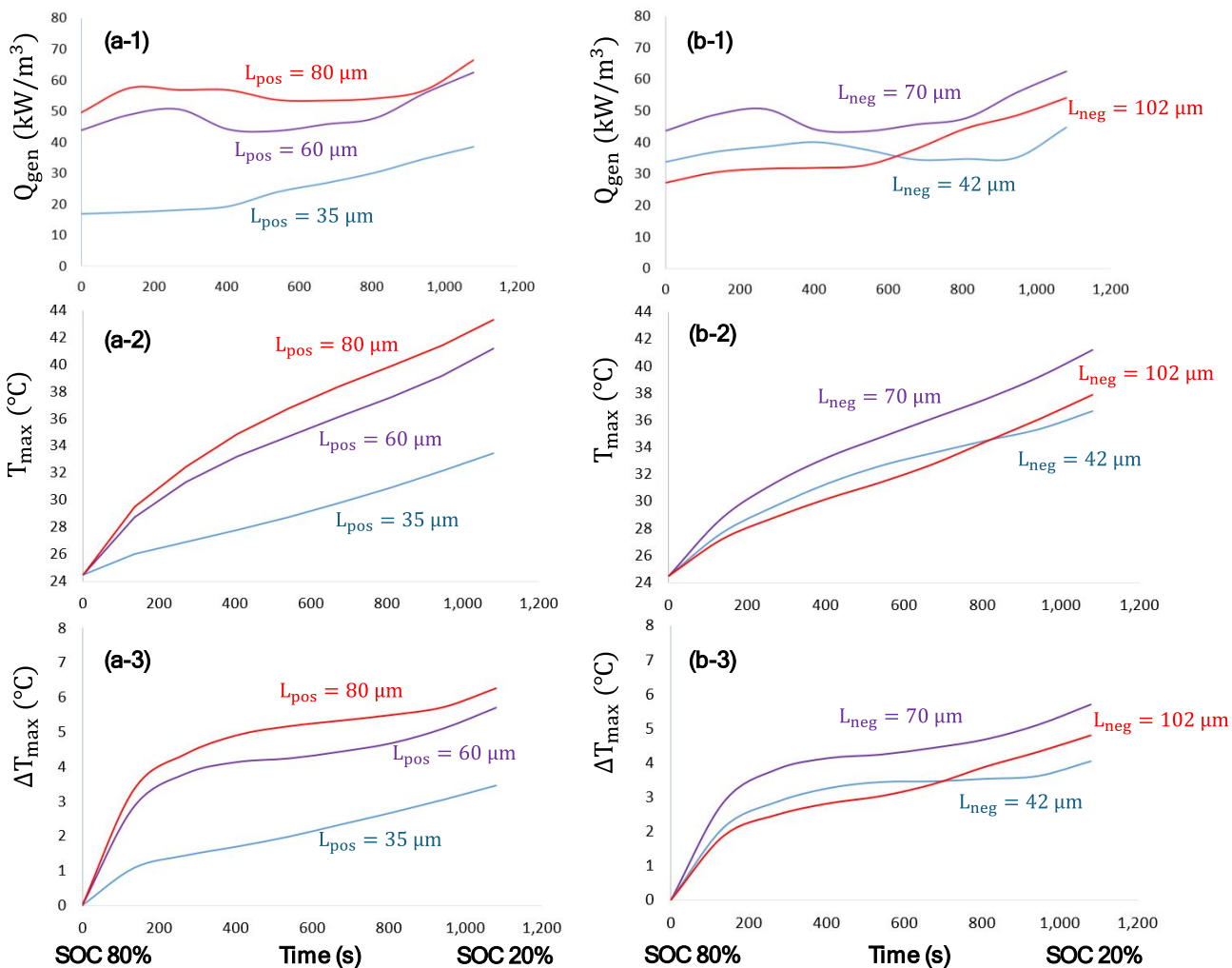


Fig. 6. Sensitivity analysis of cell thermal performance with respect to changes in electrode thicknesses, showing their impact on cell volumetric heat generation rate (Q_{gen}), maximum temperature (T_{max}), and maximum spatial temperature gradient (ΔT_{max}). Effects of variations in positive electrode thickness (L_{pos}) are presented in (a-1) Q_{gen} ; (a-2) T_{max} ; and (a-3) ΔT_{max} , and in negative electrode thickness (L_{neg}) are presented in (b-1) Q_{gen} ; (b-2) T_{max} ; and (b-3) ΔT_{max} .

6.2 Effect of Electrode Solid Volume Fraction

Figure 7 illustrates the impact of variations in the solid volume fraction of the positive electrode (ϵ_s^+) and the negative electrode (ϵ_s^-) on Q_{gen} , T_{max} and ΔT_{max} . As shown in Figs. 7(a-1) and 7(b-1), the volumetric heat generation rate (Q_{gen}) exhibits a higher sensitivity to ϵ_s^+ than ϵ_s^- , again consistent with our previous multi-length-scale study¹³. The main reason behind this finding is that higher values of ϵ_s^+ correspond to a reduced volume of the electrolyte liquid phase within the positive electrode. During discharge, this restricted liquid phase accelerates the accumulation and concentration of Li-ions within the cathode domain – often termed the saturation effect – thereby increasing the internal resistance and the resulting elevated Q_{gen} . This greater volumetric heat generation



rate drives higher values for both T_{\max} and ΔT_{\max} , as detailed in Figs. 7(a-2) and 7(a-3). Given that elevated values of T_{\max} and ΔT_{\max} – such as those observed when $\epsilon_s^+ = 0.75$ – accelerate degradation in LIB cells, optimizing ϵ_s^+ is essential for maximizing battery lifespan, while acknowledging their influence in the regulation of charge conservation, ionic conductivity and Li-ions transport kinetics.

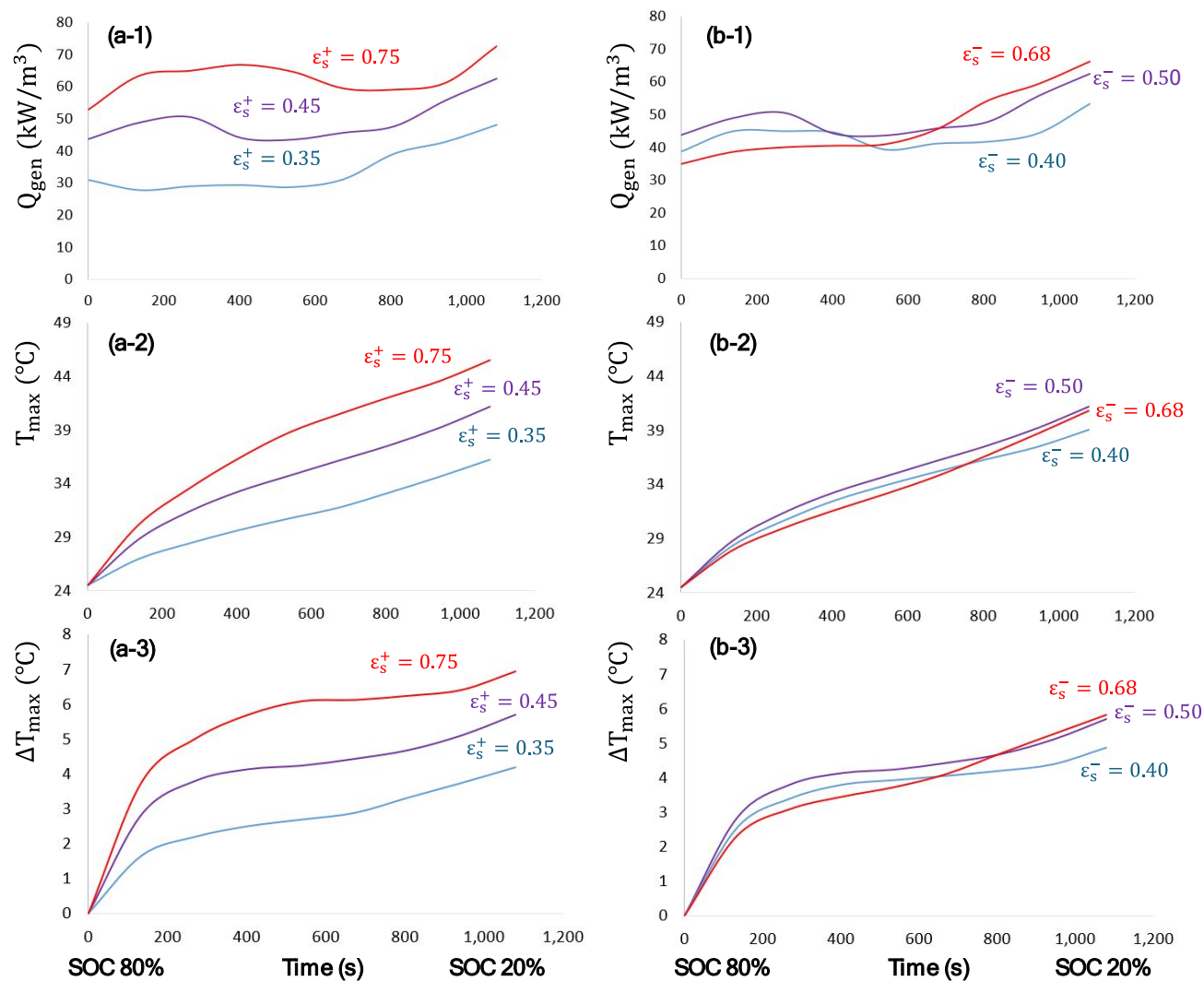
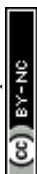


Fig. 7. Sensitivity analysis of cell thermal performance with respect to changes in electrode solid volume fraction, examining their impact on cell volumetric heat generation rate (Q_{gen}), maximum temperature (T_{\max}), and maximum spatial temperature gradient (ΔT_{\max}). Effects of variations in positive electrode solid volume fraction (ϵ_s^+) are presented in (a-1) Q_{gen} ; (a-2) T_{\max} ; and (a-3) ΔT_{\max} , and in negative electrode solid volume fraction (ϵ_s^-) are presented in (b-1) Q_{gen} ; (b-2) T_{\max} ; and (b-3) ΔT_{\max} .



6.3 Effect of Tabs Geometry

Figure 8 demonstrates the strong influence of tab width on the spatial distribution of volumetric heat generation rate by illustrating 3D contours of Q_{gen} and temperature field at the end of discharge (SOC 20%), and demonstrating how these variables respond to changes in the cell tab width. While electrode thickness and solid volume fraction impact the magnitude of volumetric heat generation rate, tab width influences its non-uniformity, current-density localization and the formation of hot spots. For the baseline case (45 mm tab width and 3.14 MA/m² of maximum current density), the results of Q_{gen} and temperature field are presented Figs. 8(b) and 8(e), respectively. Increasing the tab width to 70 mm reduces the maximum concentration of current density to 2.02 MA/m², decreasing the maximum Q_{gen} to 115 kW/m³ (Fig. 8(a)), thus lowering T_{max} and ΔT_{max} to safer levels (Fig. 8(d)). Conversely, narrowing the tabs to 30 mm intensifies localized volumetric heat generation rate to 630 kW/m³ (Fig. 8(c)) due to an increased maximum current density of 4.71 MA/m², resulting in more pronounced hot spots near the tabs (Fig. 8(f)) – a condition that compromises cell safety, reliability, and degradation uniformity. This analysis reveals that the dimensions of the cell tabs significantly impact the spatial distribution of the volumetric heat generation rate (without affecting the average magnitude), in turn causing different temperature fields and localized hot spots for a given current profile and heat-dissipation condition.

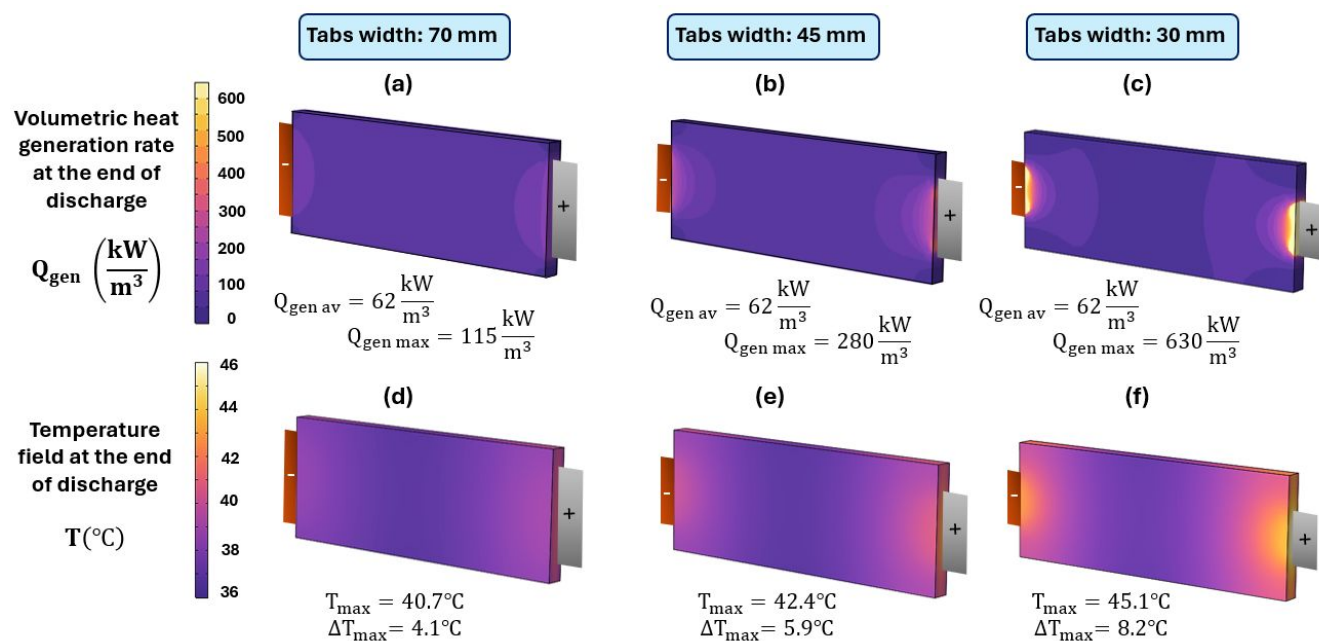


Fig. 8. Sensitivity analysis of cell thermal performance with variations in tab width at end of discharge (SOC20%). Contours of volumetric heat generation rate (Q_{gen}) are shown for tab width of (a) 70 mm; (b) 45 mm, and (c) 30 mm, detailing average ($Q_{\text{gen av}}$) and maximum ($Q_{\text{gen max}}$) values. The corresponding temperature fields, including T_{max} and ΔT_{max} , are presented for tab width of (d) 70 mm; (e) 45 mm, and (f) 30 mm.



These results clearly demonstrate that cell tab geometry is an important design parameter for controlling spatial thermal behavior in large-format pouch cells, particularly for counter-tab configuration.

7. Conclusions

In this study, we developed a three-dimensional (3D) electrochemical-thermal (ECT) coupled model for lithium-ion pouch cells with counter-tab configuration, specifically designed to achieve high-fidelity spatio-temporal temperature predictions. This model employs a hierarchical domain structure, linking sub-cell-level electrochemical processes with anisotropic cell-level thermal transport through a bi-directional exchange framework of volumetric heat generation rate and temperature field. A key novelty of this work is the introduction of a heat-generation distribution factor that accurately captures non-uniform Joule heat generation influenced by the current density concentration in the metal current collectors near the cell tabs, thereby overcoming the limitations of conventional plane-averaged projection methods. By numerically solving the governing electrochemical and energy equations, this ECT model accurately quantifies transient phenomena, including voltage responses, spatially distributed volumetric heat generation rate, and spatio-temporal temperatures.

The ECT coupled model offers essential insights for advancing pouch cell design. After being experimentally validated at different discharge C-rates, this model was used to perform a sensitivity analysis assessing the influence of electrode design parameters (thickness and solid volume fraction) and cell tab geometry on cell volumetric heat generation rate, maximum temperature, and maximum spatial temperature gradient. Results showed that the cathode design parameters affect the overall volumetric heat generation rate, while tab geometry governs its spatial distribution and the formation of localized hot spots. Thinner electrodes and lower cathode solid volume fractions are thermally beneficial, while wider tabs significantly reduce the concentration of current density, in turn decreasing localized volumetric heat generation rate and spatial temperature gradients.

By accurately identifying localized hot spots and their underlying causes, the proposed ECT model provides a powerful tool for guiding electrode design and strategies for localized heat dissipation. Overall, this work contributes to understanding and managing these complex thermal gradients in large-format pouch cells, supporting the development of more efficient, durable, and higher-performance battery systems for EV applications.



Author contributions

The manuscript was written with contributions from all authors. All authors have approved the final version of the manuscript. **Oscar A. Alvarez**: conceptualization, methodology, data acquisition, formal analysis, and original manuscript writing. **Carlos M. Da Silva**: project administration, supervision, funding acquisition, visualization, review & editing. **Cristina H. Amon**: conceptualization, formal analysis, funding acquisition, supervision, visualization, review & editing.

Data availability statement

Data will be made available on request.

Conflicts of interest

There are no conflicts to declare.

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Nomenclature

A	Area (m ²)	t	Time (s)
c _p	Specific heat capacity (J/(kgK))	t ₊	Lithium-ion transfer number
c	Lithium-ion concentration (mol/m ³)	V	Voltage (V)
D	Diffusion coefficient (m ² /s)	α	Transfer coefficient
ε _l	Liquid volume fraction	ε _s	Solid volume fraction
E	Potential (V)	η	Local over potential (V)
F	Faraday constant (96,485 C/mol)	k	Thermal conductivity (W/(mK))
h	Heat transfer coefficient (W/(m ² K))	ρ	Density (kg/m ³)
i	Current density (A/m ²)	σ	Ionic/electric conductivity (S/m)
κ	Reaction rate constant (m ^{2.5} /(mol ^{0.5} s))	S _a	Specific surface area (1/m)
L	Thickness (μm)	γ	Bruggeman exponent
T	Temperature (°C)	φ	Phase potential (V)
R	Gas constant (8.314 J/(mol K))	r _p	Particle radius of electrodes (μm)
Q _{gen}	Volumetric heat generation rate (W/m ³)	df	Distribution factor

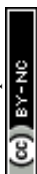


Acronyms

Lithium-ion battery	LIB	Battery Thermal Management System	BTMS
Electric vehicle	EV	Battery Energy Storage System	BESS
Root Mean Square Deviation	RMSD _T	Electrochemical-thermal	ECT
State-of-Charge	SOC	Beginning-of-life	BOL
Pseudo-Two-Dimensional	P2D	Nickel-Manganese-Cobalt	NMC
Equivalent-Circuit-Model	ECM	Lithium-ion	Li-ion

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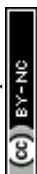
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Manuscript***A 3D Electrochemical-Thermal Coupled Model for Pouch-Type Lithium-ion Batteries with Counter-Tab Configuration*****Data availability**

Data will be made available on request.

