A One-Dimensional Stack Model for Redox Flow Battery Analysis and Operation

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Abstract: Current redox flow battery (RFB) stack models are not particularly conducive to accurate yet high-throughput studies of stack operation and design. To facilitate system-level analysis, we have developed a one-dimensional RFB stack model through the combination of a one-dimensional Newman-type cell model and a resistor-network to evaluate contributions from shunt currents within the stack. Inclusion of hydraulic losses and membrane crossover enables constrained optimization of system performance and allows users to make recommendations for operating flow rate, current densities, and cell design given a subset of electrolyte and electrode properties. Over the range of experimental conditions explored, shunt current losses remain small, but mass-transfer losses quickly become prohibitive at high current densities. Attempting to offset mass-transfer losses with high flow rates reduces system efficiency due to the increase in pressure drop through the porous electrode. The development of this stack model application, along with the availability of the source MATLAB code, allows for facile approximation of the upper limits of performance with limited empiricism. This work primarily presents a readily adaptable tool to enable researchers to perform either front-end performance estimates based on fundamental material properties or to benchmark their experimental results.

Keywords: flow battery; stack model; vanadium; performance; analysis; model; crossover; shunt current

1. Introduction

Redox flow batteries (RFBs) have been hailed as energy storage solutions that are cost-effective at large (e.g., grid) scales [1–4]. In contrast to enclosed batteries, RFBs use liquid-phase charge storage materials such that the system power, which scales with reactor area, and energy, which scales with tank volume, can be specified independently. As energy-intensive storage systems can support integration of intermittent renewables, such as solar and wind, into the grid, the scalability of these systems presents an economic opportunity. Presently, RFBs have only seen limited instances of commercial installation, and their cost-competitiveness is curbed by many factors including component cost and energy density [4–6]. Although many types of RFBs are being explored, the all-vanadium RFB (VRFB) has been most widely researched and appears to be the closest to broad commercial implementation [7,8]. Key advantages of the VRFB include the low cost of water (ca. 0.001 USD kg$^{-1}$) and sulfuric acid (ca. 0.09 USD kg$^{-1}$) and the use of a common redox-active species on both sides of the cell, enabling recovery from crossover-driven capacity fade [4,9,10]. Related RFB research areas that seek to increase energy density and/or employ inexpensive reactants, such as nonaqueous systems [11–13], and bromine- [14–16], sulfur- [17–19], and iron-based chemistries [20–23], generally focus on development of materials that overcome the undesirable VRFB performance attributes,
namely the cell voltage \( E^\theta = 1.26 \text{ V} \) and the cost of vanadium (ca. 20 USD kg\(^{-1}\)) \[1,4,24,25\]. Many of these emerging storage concepts and novel redox chemistries have demonstrated the potential to surpass the VRFB in the future; however, to date, none has shown sufficient progress to displace the VRFB as the premier RFB for large-scale energy storage \[6,17,22,23,25\].

The wealth of literature on VRFBs includes systematic component level studies on nearly every performance-driving aspect of their design and operation \[8,26–29\]. Perhaps the most-studied VRFB component is the redox-active electrolyte (RAE), as the composition of this solution impacts the hydraulic, kinetic, ohmic, and mass-transfer resistances, as well as the system stability. Past work has demonstrated improved system capacity through RAE engineering including increasing the active species concentration up to 2 M with a mixed-acid electrolyte \[30,31\] or up to 3 M with a variety of minor additives \[32\]. Individual contributions to cell losses from acid concentrations \[33,34\], the low potential (negative) electrolyte \[35–38\], and the high potential (positive) electrolyte \[39,40\] have been studied extensively, and several reviews on VRFB electrolytes describing material properties, composition, and performance have been published \[8,26,27\]. Broadly stated, typical vanadium-based RAEs have high active species concentrations (ca. 1.5 M), low to moderate viscosities (ca. 2–10 mPa s), high ionic conductivities (200–400 mS cm\(^{-1}\)) \[8\], and poor stability at elevated temperatures due to the precipitation of vanadium as V\(_2\)O\(_5\) \[32\]. In addition to these RAE-focused activities, several reports have examined membrane properties and their improvement \[41–48\], the effects of electrodes and their modification \[49–56\], flow field design \[57–61\], flow rate \[62–64\], temperature \[65,66\], and electrolyte rebalancing \[10,67\]. Although these studies are typically systematic and conclusive for the individual system of interest, the varied experimental configurations between research groups can lead to challenges when trying to combine these individual studies and broadly assess system performance and sensitivity. The field as a whole lacks an accepted control experiment comparing baseline performance of the non-focal components. Thus, attempts to compare, validate, or build upon component advances (e.g., electrode design) reported by different research groups can often be complicated by the differences in operating conditions (e.g., flow rate or electrolyte velocity, potential limitations), in cell design (e.g., flow field type and geometry, reservoir volume, electrode thickness, membrane selection), or in electrolyte conditions (e.g., temperature, composition, sparging conditions). As such, a model with which researchers can readily input their numerous system properties (e.g., electrolyte transport, kinetics, flow field geometry) would aid in the assessment of constituent component contributions and their impact on full system performance. Arguably the most accredited VRFB models have been developed through the complementary work of Shah et al. and You et al. \[68–71\], which incorporate numerous physical parameters into a 2D model of the vanadium flow cell. These cell-level models include kinetic (Butler-Volmer), conductivity, and mass-transfer losses in a similar manner to the one used here \[72,73\]. Each of these models leverage some prior developments from Newman and Tiedemann to describe the porous electrode \[74,75\]. In addition, modeling efforts have been extended beyond the individual component or cell level to stack systems several times \[24,25,76–82\], but each report employs different empirical, inaccessible descriptors of performance-driving parameters (e.g., mass-transfer correlations). Thus, although multiple models exist, they are generally difficult to use, incorporating an overwhelming number of parameters, the selection of which is always subjective even when well-informed.

Just as the myriad of studies on the VRFB have allowed for this special issue, the plethora of system parameters have inspired this particular manuscript. Here, we present first and foremost a user-friendly VRFB model with a graphical-user interface (GUI), coded in MATLAB’s (R2018a) app designer and packaged such that it can be run without MATLAB to broaden accessibility. Perhaps the greatest improvement over prior models is the ease of use and the inclusion of guidance on where additional physical processes, such as secondary reactions, could be integrated as needed. We have also incorporated algorithms for the estimation of crossover of vanadium leading to capacity fade in imperfect membranes. In the interest of preserving simplicity and accessibility of the application algorithm, each modeled component (e.g., porous electrode, membrane, manifold) employs the
simplest description that appears sufficient to capture all relevant physics. Practically, this model application presents a tool that, in its most basic form, approximates the upper performance limits of a given VRFB design. From here, second-generation users and developers can include or exclude particular effects of interest or simply examine the system performance as a function of user-specified material parameters. The major modeled components of the single cell include the porous electrodes, the membrane, and the tanks. The porous electrode model estimates cell losses from mass-transfer, kinetics, and distributed conductivity losses within the electrolyte phase using a numerical solution to a previously reported 1D polarization model [72,73]. With the growing interest in the role of mass-transfer in RFBs, we incorporate hydraulic losses through the porous electrode to offset the electrochemical performance benefits of high flow rates [58]. The cell membrane is modeled as a separator with linear concentration profiles from one side to another [83]. In the case of a passive separator (e.g., Daramic), the edge concentration is based on the instantaneous reactor concentrations, and in the case of an active membrane (e.g., Nafion 212), the concentration is based on active site saturation conditions [83–85]. The tanks are modeled as reservoirs of a fixed volume with fast homogeneous reactions reducing the solution condition to a combination of two vanadium oxidation states at any given time. To produce a full system model of the VRFB, these component models are coupled to cells with equivalent properties through a circuit network representing shunt current pathways [82,86]. Although the primary objective of this manuscript is the generation of a broadly useful modeling framework for the community, we also explore the sensitivity of predicted system performance to key parameters, such as electrolyte flow rate and the number of cells in a stack, to both illustrate their influence and validate the model results against existing literature.

2. System Models

This section describes the components of the model and their integration into the complete flow battery model, including descriptions of the porous electrodes, membrane, and shunt currents. In brief, the porous electrode model extends upon prior work to accommodate asymmetric kinetics and various states-of-charge (SoCs) [72,73]. The membrane polarization adds linearly to the cell polarization with increases in current, and the model includes a simplistic representation of both concentration- and potential-driven active-species crossover. The shunt currents resistance network is described in detail considering the system manifold geometry, and finally, the model organization within MATLAB is described.

2.1. Electrode Polarization

The electrode polarization model was adapted from past work, where it was used to estimate mass transfer coefficients as a function of operating conditions and material properties [72,73]. In that work, the systems analyzed employed model iron-based RAEs with moderately fast and near-symmetric kinetics. For the purpose of generality, we relax those assumptions enabling extension to the description of an all-vanadium system. Additionally, since the intent of this work is to provide a tool for the community, the description of this model is presented in dimensional form, in contrast to prior work which used a dimensionless form for compactness, to retain physical transparency. The domain and boundary conditions of the electrode model, describing potential as a function of position ($x$) within the porous electrode from the membrane ($x = 0$) to the current collector or flow field ($x = L_e$) is shown in Figure 1, where $L_e$ is the electrode thickness.

Within this porous electrode, only the primary electrochemical reactions and the vanadium species concentrations are considered in determining the cell potentials. Neglecting variation of local proton concentrations is reasonable in this case as the concentration is significantly higher than that of the vanadium ions (2.6 M $\text{H}^+$ vs. 1.5 M V), and the proton mobility is higher than any of the vanadium ions ($D_{\text{H}^+} (93) \gg D_{\text{IV}} (3.9) \approx D_{\text{III}} (2.4) > D_{\text{II}} (2.4) \times 10^{-6} \text{ cm}^2 \text{ s}^{-1}$ [87,88]), where $D_i$ with $i$ as a roman numerical indicates the diffusion coefficient of the vanadium ion in oxidation state $i$. Proton diffusivity is denoted as $D_{\text{H}^+}$. Additionally, omission of the protons from the kinetic expression
enables a simple translation from the VRFB to a generic redox pair. For the high potential (positive) electrode, the primary electrode reaction is described by Equation (1), in which $E^\theta$ refers to the standard state reduction potential \cite{1}.

\begin{equation}
\text{VO}_2^+ + 2H^+ + e^- \rightleftharpoons \text{VO}^{2+} + H_2O \quad E^\theta = 1.004 \text{ V vs. SHE}
\end{equation}

Similarly, for the low-potential (negative) electrode, the primary reaction is described by Equation (2) \cite{1}.

\begin{equation}
\text{V}^{3+} + e^- \rightleftharpoons \text{V}^{2+} \quad E^\theta = -0.255 \text{ V vs. SHE}
\end{equation}

![Figure 1](https://example.com/figure1.png)

**Figure 1.** Porous electrode model domain and boundary conditions. The electrode reaction described by Equation (1) for the high potential electrode or Equation (2) for the low potential electrode occurs throughout this domain.

As with past work, we begin with the Butler-Volmer Equation (3) to describe the faradaic current ($i_n$) at the surface of the porous electrode as a function of cathodic ($k_c$) and anodic ($k_a$) rate constants, surface concentrations of the oxidized ($c_{ox}^s$) and reduced ($c_{red}^s$) active species, and the potential of the solid ($\phi_1$) and liquid ($\phi_2$) phases \cite{72,89}.

\begin{equation}
\begin{split}
i_n &= n_q F k_a c_{red}^s \exp \left( \frac{\alpha n_q F (\phi_1 - \phi_2)}{RT} \right) - n_q F k_c c_{ox}^s \exp \left( -\frac{(1 - \alpha)n_q F (\phi_1 - \phi_2)}{RT} \right)
\end{split}
\end{equation}

where $R$, $T$, $F$, $\alpha$, and $n_q$ are the universal gas constant, the absolute temperature, the Faraday constant, the transfer coefficient, and the number of electrons transferred, respectively. Note that oxidation corresponds to positive current ($i_n > 0$) by this convention. Considering a pseudo-steady state condition within the porous electrode, this faradaic current can be related to the oxidized species mass-transfer coefficient ($k_{m,ox}$) and bulk concentration ($c_{ox}^b$) according to Equation (4).

\begin{equation}
\begin{split}
i_n &= -n_q F k_{m,ox} (c_{ox}^b - c_{ox}^s)
\end{split}
\end{equation}

Likewise, we can write a similar expression for the reduced species (5) with a distinct mass transfer coefficient ($k_{m,red}$) and bulk concentration ($c_{red}^b$).

\begin{equation}
\begin{split}
i_n &= n_q F k_{m,red} (c_{red}^b - c_{red}^s)
\end{split}
\end{equation}

Rearrangement of these expressions for the pseudo-steady state flux and substitution into the kinetic expression yields Equation (6), which describes the local current density as a function of kinetic and mass-transfer parameters.
we opted to code in five different correlations from the literature [58,73,91–93], as well as to provide the weighted by selectivity for each reaction, could be used in place of

\[
i_n = \frac{n_q F k_a c_{red} \exp \left( \frac{-n_q F (\varphi_1 - \varphi_2)}{R T} \right) - n_q F k_c c_{ox} \exp \left( \frac{-n_q F (\varphi_1 - \varphi_2)}{R T} \right)}{1 + \frac{k_e}{k_{m,red}} \exp \left( \frac{-n_q F (\varphi_1 - \varphi_2)}{R T} \right) + \frac{k_c}{k_{m,ox}} \exp \left( \frac{-n_q F (\varphi_1 - \varphi_2)}{R T} \right)}
\]  

(6)

Physically, this equation readily shows that in the limit of infinitely fast mass transfer, the denominator goes to 1, and the expression approaches the original Butler-Volmer Equation (3) with the surface concentration equal to the bulk concentration. For cases in which the mass-transfer coefficient is low, the local current magnitude is reduced as the denominator grows, consistent with the depletion of active material at the electrode surface. This behavior, from a description of coupled mass-transfer and kinetics within a flow-through packed bed reactor, is typical for systems with linear descriptions of mass transfer near the solid (catalyst or electrode) surface and, superficially, mimics Langmuir-Hinselwood-Hougen-Watson or Michaelis-Menten kinetics. In future work, it may be desirable to consider multiple electrode reactions, such as the effect of parasitic reactions, and/or consider performance as a function of electrode selectivity for a given reaction. In this case, a description for each electrode reaction would be derived, similar to Equation (6), and their sum, weighted by selectivity for each reaction, could be used in place of \(i_n\) in the subsequent analysis.

Relating the local reaction current density to the ionic current density (\(i_2\)) through the liquid-phase RAE requires consideration of the volume-specific electrode surface area, or electrode area per volume, \((a_e)\) as described by Equation (7), which is derived from a charge balance on a differential planar control volume within the porous electrode.

\[
di_2 = a_e i_n
\]

(7)

The RAE effective conductivity \((\kappa_{eff})\) is estimated according to the Bruggeman relation, \(\kappa_{eff} = \kappa e^{1.5}\), as a function of the RAE conductivity \((\kappa)\) and the electrode porosity \((\epsilon)\) [72]. The previously established Bruggeman factor of 1.5 originates from a derivation suggesting that tortuosity scales with porosity to the \(-0.5\) [90]. From here, the potential in the liquid phase is related to the ionic current through Equation (8).

\[
i_2 = -\kappa_{eff} \frac{d \phi_2}{dx}
\]

(8)

For the purpose of solving this problem, we apply Neumann boundary conditions at the current collector and the membrane as there is no ionic current passing through the current collector (\(i_2 \mid x = L_e = 0\)), which is an impermeable solid, and we define the current at the membrane-electrode interface to be the total cell current (\(i_2 \mid x = L_{cell} = i_c\)). For convenience, we also define the overpotential from the electrode (\(\eta\)) as the difference between the solid and liquid potentials \((\eta = \phi_1 - \phi_2)\). Since we neglect the variation in the solid potential within this model, we can combine these expressions to develop our governing differential Equation (9), which we solve numerically (MATLAB, bvp4c) to relate the applied current to the system parameters and the potential drop within the electrode (the mean value of \(\eta\)).

\[
\frac{d^2 \eta}{dx^2} = \frac{n_q F a_e k_a c_{red} \exp \left( \frac{-n_q F \eta}{R T} \right) - n_q F a_e k_c c_{ox} \exp \left( \frac{-n_q F \eta}{R T} \right)}{\kappa_{eff} \left( 1 + \frac{k_e}{k_{m,red}} \exp \left( \frac{-n_q F \eta}{R T} \right) + \frac{k_c}{k_{m,ox}} \exp \left( \frac{-n_q F \eta}{R T} \right) \right)}
\]

(9)

Mass-Transfer Coefficients

This electrode model depends on the value of the mass transfer coefficient for each redox-active species, and estimation of these values is an ongoing area of research [72,73,91]. In light of this, we opted to code in five different correlations from the literature [58,73,91–93], as well as to provide the option for a custom mass-transfer correlation of the general form shown in Equation (10), where \(\pi_i\) are dimensionless, empirical, user-defined constants, and \(Sh, Re, Sc\) are the Sherwood \((Sh = k_m d_f / D)\),
Reynolds \((Re = \rho v d_f/\mu)\), and Schmidt numbers \((Sc = \mu/\rho D)\). The default fiber diameter \((d_f)\) is 7 \(\mu\)m [94], and the viscosity \((\mu)\) and density \((\rho)\) values for both electrolytes are taken to be 5 mPa s and 1.5 g mL\(^{-1}\), respectively [8]. For simplicity, we assume that electrolyte properties are constant and do not vary as a function of SoC. In our base case for all of the included analysis, we use the correlation in Equation (11) from Barton et al., as it has specifically been demonstrated with the interdigitated flow fields [73].

\[
Sh = \pi_1 + \pi_2 Re^{0.68}Sc^{0.50}
\]

(10)

\[
Sh = 0.018Re^{0.68}Sc^{0.50}
\]

(11)

The average electrolyte velocity \((v_e)\) is calculated based on the number of channels \((n_{ch})\), the flow rate through the electrode \((Q_{cell})\), the channel length \((L_{ch})\), electrode thickness, and the electrode porosity using Equation (12).

\[
v_e = \frac{Q_{cell}}{n_{ch}L_{ch}L_e\epsilon}
\]

(12)

2.2. Membrane Effects

Incorporation of membrane properties into the model enables an analysis of voltaic losses due to the associated resistance as well as of current inefficiency due to active species crossover. The estimate of membrane area specific resistance is based on the membrane conductivity \((\kappa_{mem})\) and thickness \((L_{mem})\), and the membrane voltage drop \((\Delta V_{mem})\) is estimated according to Equation (13).

\[
\Delta V_{mem} = \frac{i_{cell}L_{mem}}{\kappa_{mem}}
\]

(13)

Since we consider the special case of the VRFB, we couple crossover to our mass balance through three homogeneous redox reactions, Equations (14–16). We assume that each of these reactions occur instantaneously, which is synonymous with having large or well-mixed reservoirs as this would enable the respective solutions to reach a composition with only two oxidation states of vanadium present. Crossover coupled to homogeneous redox reactions contributes to coulombic inefficiencies to different extents based on the ion oxidation states in solution. For example, in the case that V\(^{2+}\) crosses over to the high potential (positive) side, or VO\(^{2+}\) crosses to the low potential (negative) side, the ions will interact with those already present to achieve the intermediate oxidation state, V\(^{3+}\), as described by Equation (14). Similar self-discharge reactions are also listed in Equations (15) and (16), and the combination of these three reactions describe the self-discharge events that are assumed to occur quickly and allow each side of the battery system to maintain mass and charge balances. Additionally, inclusion of these reactions simplifies the system by allowing for the consideration of only one electrochemical reaction for each side of the cell as no more than two oxidation states of vanadium will exist for any significant length of time.

\[
V^{2+} + VO^{2+} + 2H^+ \rightleftharpoons 2V^{3+} + H_2O
\]

(14)

\[
V^{2+} + VO_2^{2+} + 2H^+ \rightleftharpoons V^{3+} + VO^{2+} + H_2O
\]

(15)

\[
V^{3+} + VO_2^{2+} + 2H^+ \rightleftharpoons 2VO^{2+} + H_2O
\]

(16)

Collectively, these equations allow coupling of the active species crossover flux \((N_{X,i})\) to Faraday’s law of electrolysis for estimation of active species concentration at each new time step during the simulation according to the Equation set (17) [83]. These equations are derived by simple consideration of the series of events occurring following crossover. For example, we describe the case in which a V\(^{2+}\) ion crosses over from the low potential (negative) side, with a reservoir volume \(V_{low}\), to the high potential (positive) side with a reservoir volume \(V_{high}\). Within the high potential (positive) reservoir the ion interacts with a VO\(^{2+}\) ion according to Equation (15) producing a V\(^{3+}\) and a VO\(^{2+}\) ion. From this
point, the V$^{3+}$ ion interacts with another VO$_2^+$ ion to form two VO$^{2+}$ ions according to Equation (16). In total, this crossover of a single V$^{2+}$ ion produced three VO$^{2+}$ ions and consumed two VO$_2^+$ ions, and this is reflected by the coefficients on the crossover flux of V$^{2+}$ in Equation set (17). Note that the relative impact of crossover on the composition profile depends on the ratio of the cell area (A) to the reservoir volume (V$_r$). Also note that if one were to introduce the effect of a side reaction with a rate specified on a volumetric basis, such as vanadium precipitation [32], then that tank reaction could be incorporated within the expressions in Equation set (17).

$$\frac{dc_{II}}{dt} = \frac{A_{V^{II}}}{V_{cell}} \sum \left( \frac{c_{II}}{F} - N_{X,III} - N_{X,IV} - 2N_{X,V} \right)$$

$$\frac{dc_{III}}{dt} = \frac{A_{V^{III}}}{V_{cell}} \sum \left( - \frac{c_{III}}{F} - N_{X,III} + 2N_{X,IV} + 3N_{X,V} \right)$$

$$\frac{dc_{IV}}{dt} = \frac{A_{V^{IV}}}{V_{ membrane}} \sum \left( - \frac{c_{IV}}{F} - N_{X,IV} + 2N_{X,III} + 3N_{X,II} \right)$$

$$\frac{dc_{V}}{dt} = \frac{A_{V^{V}}}{V_{ membrane}} \sum \left( \frac{c_{V}}{F} - N_{X,V} - N_{X,III} - 2N_{X,II} \right)$$

Active species crossover is accounted for based on literature values of V$^{2+}$, V$^{3+}$, VO$^{2+}$, and VO$_2^+$ permeabilities ($K_i$) through Nafion 117 [41]. The flux through the membrane may be modeled according to Equation (18) whether an active separator/membrane (active sites and surface charge play a role) or passive separator (porous separator without additional chemical functionality or interactions) [83]. Arguably, this treatment is incomplete as we neglect effects from supporting species (e.g., H$^+$, H$_2$O, etc.), but it is practical and provides a qualitative approximation of the crossover flux for each species $i$ ($N_{X,i}$) as a function of the primary system properties, such as ion valence ($z_i$), membrane thickness, ion concentration, and the potential gradient across the membrane, which scales with the current density.

$$N_{X,i} = -K_i \frac{dc_i}{dx} + \frac{z_i K_i F c_i}{\kappa_{mem} R T} i_{cell}$$

(18)

This expression has been solved for the case of linear concentration profiles in the membrane, and the solution for non-zero currents is described by Equation (19), where $c_{i,0}$ is the concentration of species $i$ at the membrane surface [83]. For open-circuit conditions, Equation (19) is singular, and the simple diffusion problem is solved instead, resulting in Equation (20). These expressions are evaluated for each individual cell and assumed to be uniform over the cell area. Additionally, the flux is taken to be positive for each species as it leaves its native side of the cell, consistent with Equation set (17). Estimation of $c_{i,0}$ requires a priori knowledge of the membrane behavior. In the case of an active separator/membrane (e.g., Nafion), some condition of active site saturation may be assumed based on the solution conditions, 1.5 M V here [83–85]. In the case of a passive separator with a non-specific selectivity (e.g., Daramic), the concentration may be taken as a planar average of the active species concentration using the separator porosity ($c_{i,0} = c_i e_{mem}$) as its actual liquid-phase concentration is effectively depressed by the inclusion of solid volume.

$$N_{X,i} = \frac{z_i F K_i c_{i,0} i_{cell}}{\kappa_{mem} R T} \frac{1 - e^{-z_i F c_i e_{mem} / \kappa_{mem} R T}}{z_i F c_{i,0} e_{mem} R T i_{cell}}$$

(19)

$$N_{X,i} = \frac{K_i c_{i,0}}{L_{mem}}$$

(20)

2.3. Shunt Resistance Network

Shunt currents in multi-cell electrochemical systems are driven by large voltage differences across continuous paths of ionically conductive solutions [77,78,81,86]. Practically, these ion pathways reduce the system efficiency and can even lead to corrosion at the electrode edges [86], a failure mode not considered here. The manifestation of shunt currents within full RFB systems allows cell-to-cell
variations in current density and computationally complicates the polarization model as multiple cells must be solved simultaneously with the resistor network depicted in Figure 2.

\[ I_{cell}(n+1) - I_{cell}(n) - 2I_{p,high}(n) - 2I_{p,low}(n + 1) = 0 \]
\[ R_{p,high}I_{p,high}(n) - R_{p,high}I_{p,high}(n + 1) - R_{m,high}I_{m,high}(n) = E_{cell}(n + 1) \]
\[ R_{p,low}I_{p,low}(n) - R_{p,low}I_{p,low}(n + 1) - R_{m,low}I_{m,low}(n) = E_{cell}(n) \]
\[ I_{p,high}(n + 1) + I_{m,high}(n + 1) - I_{m,high}(n) = 0 \]
\[ I_{p,low}(n + 1) + I_{m,low}(n + 1) - I_{m,low}(n) = 0 \]  

Additionally, consideration of the end nodes requires five additional equations, Equation set (22), for a full description of the system, where \( I_{tot} \) is the total current applied through the stack.

\[ I_{cell}(1) - I_{tot} - 2I_{p,low}(1) = 0 \]
\[ I_{p,low}(1) + I_{m,low}(1) = 0 \]
\[ I_{p,high}(1) + I_{m,high}(1) = 0 \]
\[ I_{p,low}(N) - I_{m,low}(N - 1) = 0 \]
\[ I_{p,high}(N) - I_{m,high}(N - 1) = 0 \]  

Although these sets of equations are linear as presented, the cell voltage is a nonlinear function of current density. Past literature has given estimates of these shunt currents, and in all cases, the shunt current magnitudes are small relative to the total system current (< 2%) [77,78,81]. As a result, the variations in cell voltage are minimal. In order to minimize evaluations of the nonlinear cell voltage function and accelerate computation, we linearize the cell voltage about the applied current density using a resistance \( r_{cell} \) to the incremental change in cell current due to the shunt currents. This approximation is shown in Equation (23), and the value of \( r_{cell} \) is evaluated numerically using two
function evaluations according to Equation (24), which come from the definition of the derivative and require a model decision in the selection of the current step ($\Delta I$) for the evaluation of this derivative.

$$E_{\text{cell}} \approx E_{\text{cell}}|_{I=I_{\text{tot}}} + r_{\text{cell}} \cdot (I_{\text{cell}} - I_{\text{tot}}) \quad (23)$$

$$r_{\text{cell}} = \frac{E_{\text{cell}}|_{I=I_{\text{tot}}+\Delta I} - E_{\text{cell}}|_{I=I_{\text{tot}}}}{\Delta I} \quad (24)$$

If we consider an isolated loop in Figure 2 as current passes from the electrode of a cell through a port, then a single manifold segment, then back through a port to the electrode of the next cell, we can obtain a coarse estimate of the port current magnitude based on Equation (25). Considering that there are two ports (inlet and outlet) connected to each half cell, we take double this value as our current step ($\Delta I \approx \pm 2I_{\text{port}}$). For context, in the base case, where we cycle the cell at 100 A, this value is on the order of 3 mA.

$$I_{\text{port}} \approx \frac{E_0}{2 \left( R_{\text{port,low}} + R_{\text{port,high}} \right) + \frac{R_{\text{man,low}} + R_{\text{man,high}}}{2}} \quad (25)$$

2.4. Hydraulic Losses

Electrode permeability ($K$) is estimated using the Carman-Kozeny Equation (26) similar to past work [95], but the user may opt to input a measured permeability as well. For all the results presented within this manuscript, we use a Carmen-Kozeny constant ($k_{\text{CK}}$) value of 4, which corresponds to randomly aligned fibers [95]. We note here that in a real system, the permeability and mass-transfer resistance are inherently related; however, the mass-transfer model incorporated here does not fully capture this behavior, ignoring the effect of $k_{\text{CK}}$ on the mass-transfer losses.

$$K = \frac{d_f^2 \epsilon^3}{16 k_{\text{CK}} (1 - \epsilon)^2} \quad (26)$$

Following front-end evaluation of the permeability, the pressure drop ($\Delta P$) across the porous electrode is estimated using an analytical expression (27) from past work for the interdigitated flow field based on the volumetric flow rate through an individual cell ($Q_{\text{cell}}$) [58,96]. Intermediate steps include the calculation of the channel hydraulic diameter ($d_h$) in Equation (28) based on the channel width ($w_{\text{ch}}$) and height ($h_{\text{ch}}$), as well as a dimensionless geometry-informed permeability factor ($\xi$) described in Equation (29) that captures the combined effects of permeability and design parameters, such as the channel length ($L_{\text{ch}}$) and rib width ($w_{\text{rib}}$).

$$\Delta P = \left( \frac{32 \mu Q_{\text{cell}} L_{\text{ch}}}{N_{\text{ch}} w_{\text{ch}} h_{\text{ch}} d_h^2} \right) \left( 1 + \frac{2 + 2 \cosh \xi}{\xi \sinh \xi} \right) \quad (27)$$

$$d_h = \frac{4w_{\text{ch}} h_{\text{ch}}}{2(w_{\text{ch}} + h_{\text{ch}})} \quad (28)$$

$$\xi = \frac{128 L_{\text{ch}}^2 K L_e}{d_h^2 (L_e + w_{\text{rib}} + w_{\text{ch}})(w_{\text{ch}} h_{\text{ch}})} \quad (29)$$

2.5. Model Framework

To increase accessibility, the stack model was written within the MATLAB app designer as this produces a simple graphical user interface (GUI) for visualizing how varying different inputs can impact system performance. Each of the four tabs of the GUI are depicted in Figure 3. Where possible, system inputs were divided between the tabs according to topic: (1) material properties, (2) system properties, (3) simulation, and (4) manifold, membrane saturation, and polarization.
Here, we note the general overview of system data flow, and the application parameters and functions are classified within Figure 4 to supplement this description. Figure 4 gives a brief list of all the system parameters as well as a short description of all the application functions. The simulation loads in all the user inputs (e.g., electrode porosity) from the GUI, and any changes to these values result in immediate recalculation of the continuously calculated parameters (e.g., mass-transfer coefficients). Toggled options generally affect the interpretation of user inputs (e.g., membrane behavior) or specific options for the conditions of cycling (e.g., limiting the number of cycles simulated). These functions are typically low-complexity; however, the options themselves can have a significant impact on the results of a given simulation. In contrast to this, activated operations may involve lengthier computing time (e.g., simulating cycling) or operations meant to facilitate user experience (e.g., loading references). For example, in order to simulate galvanostatic cycling, the user sets all system properties to appropriate values across all the tabs, then clicks “Run” on the “Simulation” tab to start cycling the VRFB. While running, a popup window will display progress towards the simulation time completed. For the first use, we suggest using the “Estimate Time” button before running, which evaluates the user’s systems speed at solving an ordinary differential equation using MATLAB’s “bench” function and scales this based on the number of time steps needed. After loading the system properties, the application builds the matrix describing the shunt current network and initializes the cell potential. Then, it integrates forward in time from this initial point, evaluating the cell voltages and updating the value of $r_{cell}$ within the shunt current matrix, then solving for the cell, port, and manifold currents at each time step. Additionally, when a half-cycle (charge or discharge) finishes, the element capturing the applied current within the shunt current matrix descriptor must also be updated (as the sign changes). After this solution, the active species concentrations are updated according to Equation (17). Finally, at each time step, the currents and cell voltages relative to ground are all stored in immediate recalculation of the continuously calculated parameters (e.g., mass-transfer coefficients). 

Figure 3. Model graphical user interface (GUI). (a) Material properties, such as electrode properties, bulk redox-active electrolyte (RAE) properties, and membrane characteristics. (b) System properties, such as flow field geometry and operating conditions. (c) Simulation options, such as time step and output file names, as well as simulation outputs, such as pressure drop and voltage history. (d) Assorted interface components, such as manifold geometry, membrane behavior, a cell polarization tool, and a place for user-input notes.
as data accessible by opening the exported Microsoft Excel file or MATLAB workspace. The names of these files are taken from user inputs on the “Simulation” tab of this function. In the case that a single cell is analyzed, this employs a slightly shortened algorithm as shunt currents are not present. After completion of this forward walk through time, a few variables are plotted within the application, specifically, the species concentrations and the system total and cell average voltages are plotted as a function of time. All of the employed code as well as the free-standing application are included in the supporting information (SI) for complete documentation and future implementation. The selected default values of all system parameters are listed in Appendix A in Table A1 along with clarifying illustrations of system geometry in Figure A1.

3. Results and Discussion

Within this section, a short demonstration of the single-cell model results is shown to validate appropriate behavior from our descriptors of the core physics. Then, we illustrate the impact of active species crossover, which does not inherently produce capacity fade, but does impact the coulombic efficiency and the accessible capacity. Finally, we present a few parametric studies and discuss the implications of these on the overall system performance. The absolute values of the performance metrics presented here are based on the default values selected (listed in Table A1) and will change with differing inputs but the trends described are robust. As the model stands, our results are generally bounded by previous reports of system performance, whether focused on the study of individual components or the overall system, but subject to the user input values.
3.1. Single Cell Performance

The performance of a single cell is examined in the absence of crossover to illustrate the performance of the electrode polarization model and examine sensitivity to some core material parameters, such as the mass-transfer coefficients and kinetic rate constants. To start, Figure 5a shows the impact of mass transfer coefficient on the individual cell performance. The figure itself shows polarization curves at 50% SoC using three different mass transfer coefficient correlations: Schmal et al. [92], Barton et al. [73], and You et al. [91]. While the actual conditions of the cell (1.5 M V, 2.12 L min\(^{-1}\)) are the same, there is a large variation in the predicted mass transfer coefficients from these empirical correlations. In recognition of the large variation and its impact on estimates of cell performance, we incorporated several empirical correlations [58,73,91–93] directly within the application as well as the option to enter a user-specified correlation in the format of Equation (10). The more recent correlations predict smaller mass-transfer coefficients, and this should be kept in mind throughout the subsequent analysis in which the correlation from Barton et al. [73] is selected as it has been demonstrated specifically with interdigitated flow fields. The frequently cited correlation from Schmal et al. was developed specifically for mass-transfer to an individual fiber, and the experimental setup used in 1985 may not be an accurate representation of the conditions achievable within current-generation VRFBs [92]. In this prior work, the commonly cited correlation was developed for carefully aligned fibers in a flow-through geometry, but comparison of their reactor bed with a modern RFB reveals a few quantifiable differences: typical fiber electrodes are not fully aligned with the direction of flow, interdigitated flow fields have significant flow orthogonal to the electrode plane, carbon paper porosity is typically lower than in the felt employed (ca. 80% vs. 96%), and the fibers are typically smaller than those found in the felt (ca. 7 \(\mu\)m vs. 15 \(\mu\)m) [92]. Perhaps the most interesting aspect of this particular figure is the indication that there is room for improvement in the characterization of mass transfer for RFBs. All three of these prior publications considered \(d_f\) to be the relevant length scale, but it may be important to consider separately the electrolyte path length, electrode thickness, and possibly an independent descriptor for the pore diameter or other details of the electrode microstructure. In particular, the path length within the porous electrode varies significantly with flow field design and experimental setup [73,91,92]. All three of the polarization curves in Figure 5a use literature values for vanadium reaction kinetics [88], which happen to use the same value for the anodic and cathodic reaction directions although different values are used for the high potential and the low potential reactions. As a result, these curves are symmetric about the open-circuit voltage \(V_{OC}\) as the difference in individual mass transfer coefficients is nearly imperceptible at the cell level. In Figure 5b, we manipulate these rate constants to illustrate the impact that a 10-fold increase or decrease of the kinetic rate constants would have on the cell polarization. This is meant to emphasize that if the kinetic rates previously demonstrated can be achieved within a scaled cell, there is little incentive to catalyze the reactions further as an order-of-magnitude increase produces only marginal improvement in the polarization. Performance-driven research would likely be better served furthering membranes, flow fields, or electrode architecture. Conversely, if those rate constants cannot be achieved, their increase could significantly improve cell performance. Figure 5c shows a single (second cycle) cell discharge for three different current densities at a fixed flow rate. Due to the moderate value of the mass-transfer coefficient estimated, the cell reaches 81% of the theoretical capacity (20.1 Ah L\(^{-1}\)) at low current densities, and as this increased to a more moderate value of 300 mA cm\(^{-2}\), this drops sharply to 30% of the theoretical. For comparison, Jiang et al. report accessing 72% of the theoretical capacity at 40 mA cm\(^{-2}\), and decreasing to 25% at 320 mA cm\(^{-2}\) [97]. Figure 5d illustrates the influence of current density and flow rate on the cell voltaic efficiency and accessed capacity. The voltaic efficiency (black, left) decreases from ca. 95% to 88% as the current density is increased with a scaled flow rate from 100 mA cm\(^{-2}\) to 300 mA cm\(^{-2}\) at 2.12 L min\(^{-1}\) and 6.36 L min\(^{-1}\), respectively. This particular area-specific flow rate (2 mL min\(^{-1}\) cm\(^{-2}\)) is comparable to reported and optimized values within the field [58,62]. If scaled in this manner for this 1000 cm\(^2\) cell, the accessed capacity falls from 88% to 67% of the theoretical in contrast to Figure 5c, where flow rate was held...
The increase in flow rate increases the accessed capacity as well as the voltaic efficiency due to the increases in the mass-transfer coefficients. Practically, the pumping losses also increase linearly with flow rate from 16 kPa to 48 kPa for 2.12 L min\(^{-1}\) to 6.36 L min\(^{-1}\). Since the pressure drop scales linearly with flow rate, the pump work scales with its square, and in terms of pump power, this corresponds to an increase from 1.6 W to 15 W \((3^2 \times)\) with a 70% pump efficiency. Although this may seem a subtle point, it illustrates that the hydraulic losses within a flow battery grow quickly as flow rate is increased in an attempt to overcome mass-transfer resistance. This relationship poses a dynamic problem for the efficient operation of VRFBs as the SoC and system load change. High-power discharge requires some minimum flow rate, which negatively impacts the efficiency.

**Figure 5.** Single cell performance. Discharge (○) and charge (△) polarization at 50% state-of-charge (SoC) and 1.5 M total active material concentration and a fixed flow rate of 2.12 L min\(^{-1}\) using (a) three different mass-transfer correlations [73,91,92] and (b) three different sets of kinetic parameters, where literature (black) kinetic values are compared to a 10× increase (red) or decrease (blue). (c) Galvanostatic discharge curves for three different current densities at 2.12 L min\(^{-1}\). (d) Single cell voltaic efficiencies (black, left) and discharge capacities (blue, right) as a function of current density and flow rate.
3.2. Membrane Crossover

Figure 6 shows the impact of crossover and separator behavior (active or passive) on the performance and capacity of a single-cell. The accessed capacities on charge and discharge are shown in Figure 6a for cases with and without crossover enabled for a single 1000 cm$^2$ cell operating at 100 mA cm$^{-2}$ with 2.5 L tanks and flow rate of 2.12 L min$^{-1}$. The relatively small tank size was selected to keep the simulation time short and approximate common experimental energy-to-power ratios. Permeabilities are listed in Table A1 [41]. Crossover is easily removed by setting the membrane permeability to each active species to 0. The theoretical half-cycle for this configuration is 1 h, and over the 50 cycles shown, the active membranes (Nafion 212 and 117) show low capacity fade. In the case of these cation-exchange membranes, the varied solubilities (listed in Table A1 [84]), or saturation concentrations, of vanadium within the membrane bias crossover in one direction adding to the imbalance of total vanadium on either side. For the conditions shown, the Nafion coulombic efficiency remains high, and the capacity fade is less than many reports of experimental results [97–99], likely due to the omission of supporting species crossover. Here, the capacity fade over 50 cycles was 10.7% and 1.9% of the initial accessed capacity for the Nafion 212 and Nafion 117, respectively. For comparison, Jiang et al. show a capacity loss of over 50% and 20% for Nafion 212 and Nafion 117 over 50 cycles [97]. Figure 6b shows the self-discharge from 50% SoC at open-circuit conditions with the same flow rate and tank volume. Practically, this long self-discharge time for a system with an energy-to-power ratio of 1 h is promising for the field. If this low permeability to active species is genuinely achievable, losses due to crossover and self-discharge over the course of a cycle should remain small, and the selectivity of these membranes may be sufficient to satisfy performance metrics for coulombic efficiency. If other losses were included, such as side-reactions, these may present convoluting effects on the system performance, for example, shifting of the average vanadium oxidation state, and may further reduce the system capacity. As it stands, omission of supporting species crossover from the model likely causes the underestimation of capacity fade observed here. The trends are qualitatively correct. We also attempted to include a comparison of a passive separator with a MacMullin number of 4.5 and thickness of 25 µm, similar to Celgard 2500 [100]. The results are not shown here as the capacity fade was rapid enough to prevent meaningful discussion. We expect that the error in the model is greater for the high fluxes experienced with a passive separator due to the omission of supporting species transport.

![Figure 6. Single cell crossover impact. (a) Charge (△) and discharge (▽) capacity as a function of cycle number at 100 mA cm$^{-2}$. Inclusion (black, red) and omission (blue) of crossover clearly indicate its impact on capacity fade with stable current efficiency. (b) Change in active species concentrations as a function of time at open-circuit conditions due to the rate of crossover. Electrolytes contained 1.5 M V initially.](image)
3.3. System Performance

This section demonstrates the impact of some system parameters, such as the number of cells in the stack, on system performance metrics, such as power losses. First, we present resulting shunt currents during discharge for a range of system stack sizes. For the sake of a clear comparison, tank volume and total flow rates are scaled with the number of cells (10 L and 2.12 L min⁻¹ per cell) to maintain a constant energy-to-power ratio (4 h at 100 mA cm⁻²) and electrolyte velocity. Figure 7a shows the shunt currents within the ports and manifold as a function of position within the stack. These current magnitudes and trends are similar to past work [78,82]. The edge conditions force the manifold current to match the port current, and this is directly observable at cell N for each stack size. At cell 1, the port and manifold currents are equivalent in magnitude and direction, but opposite in sign merely due to the convention chosen in Figure 2. The applied current, 100 A, is much larger than the port and manifold currents, and as a result, the impact of these shunt currents on the cell voltage is minimal as an individual cell voltage is relatively insensitive to a 0.1% change (100 mA vs. 100 A) in the current passing through it. Practically, the port currents may be considered leakage currents from individual cells, and this leakage is summed within the manifold to produce a maximum manifold current near the center of the stack. The port currents are the greatest near the end of the stack and they are minimal near the middle of the stack. Physically, this phenomenon arises due to the fact that each successive cell drives current through the port in the opposite direction of the previous cell. However, at the ends the ports only see one cell thus the leakage (port) currents can grow to a greater value. Figure 7b shows the increase in mean discharge power (left) as a function of stack size alongside the power lost to shunt currents and pump work (right). Across the cases explored, pump work and shunt losses grow significantly with the stack size. However, the relative magnitude of these losses remains small. The shunt current magnitude is consistent with prior reports, both in terms of the relative scale of these currents and the dependence on position within the stack [78,81]. Practically, quantitative comparison of these values is difficult as simple variations in system geometry can significantly affect the results.

**Figure 7.** Stack losses during discharge at as a function of the number of cells when cycled at 100 mA cm⁻² with a flow rate of 2.12 L min⁻¹ cell⁻¹ and 1.5 M V. (a) Mean port (△) and manifold (○) currents as a function of position within the stack. (b) Mean stack discharge power (○, left) alongside energy losses due to shunt currents (△, right) and pump work (□, right).

Figure 8 illustrates the impact of flow rate on performance of a 35-cell stack. Figure 8a shows the steady increase in both capacity (left) and energy (right) with increases in flow rate. Increases in
flow rate correspond to an increase in the mass-transfer coefficient, thus increasing system energy both by reducing the limiting reactant concentration and by increasing the cell voltage during discharge. Figure 8b shows the impact of flow rate on the average discharge power, pump power, and roundtrip energy efficiency. At low flow rates, increasing the flow rate gives rise to a substantial increase in cell power, but at high flow rates, the pumping losses quickly grow to detract from the overall system efficiency. This loss is emphasized by the difference between the blue markers, which show the roundtrip efficiency with and without accounting for the pumping losses. It is reassuring to see that the heuristically selected flow rate of 2 mL min\(^{-1}\) per cm\(^2\) of active area is near the point of maximum efficiency. In this case, the system operates at slightly higher efficiency at 1.5 mL min\(^{-2}\) per cm\(^2\) of active area, so if we consider roundtrip energy efficiency to be our primary variable to maximize and constrain the system to an invariant flow rate, then this would be the optimal condition. However, both logic and prior art suggest that to maximize efficiency, a variable flow rate should be employed \([101,102]\). The system is generally insensitive to mass-transfer coefficient while the reactant is being consumed from an enriched state (i.e., discharging at high SoC or charging at low SoC), thus operation at low or moderate flow rates is optimal here. When the reactant becomes depleted, it then becomes worthwhile to increase the flow rate, enhancing the mass-transfer coefficient and enabling greater accessible capacity. We also note the slope of the efficiency curve to either side of the flow rate corresponding to maximum efficiency. To the left of the maximum, decreases in flow rate produce a sharp decrease in efficiency, whereas to the right, the decrease is slower, suggesting that in the case of uncertainty, erring towards higher flow rates may be favorable. Increasing current density increases the flow rate of maximum efficiency due to the increase in mass-transfer losses (data not shown). Additionally, higher current densities reduce the maximum efficiency due to an increase in all sources of losses. Figure 9 examines the degree to which current density influences the system performance.

Figure 8a shows the impact of current density on the cell-averaged discharge curve at fixed flow rate. As the current increases, the cell voltage and accessed capacity decrease in a near linear fashion as shown in Figure 9b. The accessed capacity decreases from 87% to 30% of the theoretical capacity (20.1 Ah L\(^{-1}\)) as the current density increases from 50 mA cm\(^{-2}\) to 300 mA cm\(^{-2}\). The voltage efficiency also decreases from 96% to 81% with this increase in current density. A thorough comparison
with the literature reveals that these efficiencies are definitively overestimated at low current densities, but within the range of reported values at high current densities, as results are widely varied in this regime [103–107]. For comparison, Zhou et al. report a voltaic efficiency of 90% and 67% at 80 mA cm$^{-2}$ and 320 mA cm$^{-2}$, respectively, but the authors use Nafion 115, which should have a resistance of 2.5× the value used here, representing Nafion 212 [103]. Additionally, this same work reports similar accessed capacities at both low and high current densities (82% and 20% of the theoretical at 80 mA cm$^{-2}$ and 320 mA cm$^{-2}$, respectively) [103]. Wang et al. report voltaic efficiencies ranging from 92% to 67% at 40 mA cm$^{-2}$ and 200 mA cm$^{-2}$, respectively [104]. Recent demonstration and analysis of controlled-power cycling of a system with multiple vanadium flow battery stacks at 50 kW to 200 kW showed voltage efficiencies of 91% to 80% [105]. In this case, the membrane was not reported, and the estimation of the average current density is unclear. The overestimation of the voltage efficiency at low current density suggests that the kinetic losses may be underestimated and, although these kinetic parameters have been used in past work, their true values likely vary greatly between systems, especially as a function of electrode selection and pretreatment. At high current densities, differences within the literature are more likely explained by variations in the mass-transfer coefficient between systems. For the default conditions employed, this model estimates a limiting current density of a little over 400 mA cm$^{-2}$ as shown in Figure 5a, but within the literature, one can find limiting current densities for similar electrolyte compositions as low as 60 mA cm$^{-2}$ or greater than 2600 mA cm$^{-2}$ [106,107]. These differences arise as a result of the large variations in experimental setup, and serve as motivation for our incorporation of multiple mass transfer correlations as detailed in Section 2.1. As the model stands, these results are generally bound by previous reports of system performance and emphasize the sharp decrease in efficiency and accessed capacity as the limiting current density is approached.

![Figure 9](image-url)

**Figure 9.** Performance of the 35-cell stack over a range of current densities at 74.2 L min$^{-1}$. (a) Discharge curves and (b) accessed capacity (○, black, left) and voltage efficiency (△, blue, right) as a function of applied current density.

### 4. Conclusions

We have discussed the development and initial validation of a RFB model packaged as a MATLAB application (see Supplementary Materials), which can be used to assess the upper performance limits of a particular system, here a VRFB. This application advances the field as the model allows for user-selected material properties. Specifically, we chose to emphasize the large variation in predicted mass-transfer coefficients through the incorporation of multiple correlations. Complete, irrefutable descriptions of this parameter as a function of all system parameters remains elusive. Additionally, this model specifically examines performance with an interdigitated flow field and represents the
first implementation of this polarization model at the stack scale, by treating the porous electrode as a well-mixed flow-through reactor. Although the default parameters may underestimate the kinetic or ohmic losses, the application allows for user variation of these and any other model parameters, to evaluate performance under conditions of interest. As it stands, this model is a flexible platform, which can be used or modified to address specific questions within the field. For example, this model could be used to estimate short-time-scale performance of other unimolecular redox chemistries (i.e., polarization) by setting the permeabilities to 0, offering a simple side-by-side comparison of a novel chemistry to the VRFB. Some clear steps towards the continued advancement of this model would be the inclusion of economic parameters, additional decay pathways and side reactions, and optimization options. The location within the code to incorporate these additional physical processes has been indicated within the text. This model would also benefit from scientific consensus on the values of different physical parameters, such as the kinetic and mass-transfer coefficients. A potential pathway toward resolving this challenge would be agreement on a common flow cell architecture, preferably small-scale, low cost and easy to manufacture, and the establishment of standardized testing protocols with readily available baseline materials.

Looking forward with respect to the model developed here, we note its limitations in order to present opportunities for improvement in subsequent analysis. First, the model expects and handles the system as if the conversion-per-pass is low. In the case of high conversion, concentration distributions of the bulk active species are significantly altered as the fluid flows along its path through the porous electrode. In such a case, the path should be segmented, and each segment could be solved as if well-mixed, similar to here. Second, the linearization of the cell resistance assumes that the shunt currents are small relative to the applied current. As a result, with low-resistance ports and manifold segments, this model will likely over-predict the system voltage as it underestimates the impact of current leakage on the individual cell potential. Additionally, at very low applied currents, the model may over-predict the impact of leaking current on the cell voltage and will likely underestimate the system voltage. Third, many additional factors affecting performance are not considered, and significant contributions from any of these factors not included within the model can lead to gross errors in performance estimates. Addressing one or all of these aforementioned limitations will further improve model fidelity and we hope the community builds upon this first release.

Supplementary Materials: The following are available online at http://www.mdpi.com/2313-0105/5/1/25/s1, Free standing application installer file: Stack_Simulator_Installer.exe, and the simulator code files for use in MATLAB R2018a: Cell_simulation_gui_v22.mlapp is the main file, and model_electrode_loss_bvp4c.m incorporates the 1D polarization model.

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### Symbol definitions.

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</tr>
<tr>
<td>$d_p$</td>
<td>port geometric diameter</td>
<td>L</td>
<td>8 mm [82]</td>
</tr>
<tr>
<td>$d_{mi}$</td>
<td>manifold geometric diameter</td>
<td>L</td>
<td>10 mm [82]</td>
</tr>
<tr>
<td>$L_p$</td>
<td>port length</td>
<td>L</td>
<td>100 mm [82]</td>
</tr>
<tr>
<td>$L_{mi}$</td>
<td>manifold interport distance</td>
<td>L</td>
<td>6 mm [82]</td>
</tr>
<tr>
<td>$V_{res}$</td>
<td>reservoir volume</td>
<td>L³</td>
<td>350 L</td>
</tr>
<tr>
<td>$V_{OC}$</td>
<td>open-circuit voltage</td>
<td>M L² I⁻¹ t⁻³</td>
<td>1.4 V [109]</td>
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<tr>
<td>$E_{low}$</td>
<td>lower voltage limit for cycling</td>
<td>M L² I⁻¹ t⁻³</td>
<td>5.5</td>
</tr>
<tr>
<td>$E_{high}$</td>
<td>upper voltage limit for cycling</td>
<td>M L² I⁻¹ t⁻³</td>
<td>8.5</td>
</tr>
<tr>
<td>$T$</td>
<td>temperature</td>
<td>T</td>
<td>22 °C</td>
</tr>
<tr>
<td>$Sc$</td>
<td>Schmidt number</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>$Pe$</td>
<td>Pélet number</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>$D_i$</td>
<td>diffusivity of species $i$</td>
<td>L² t⁻¹</td>
<td>m² s⁻¹</td>
</tr>
<tr>
<td>$v_i$</td>
<td>electrolyte average velocity</td>
<td>L t⁻¹</td>
<td>m s⁻¹</td>
</tr>
<tr>
<td>$Q$</td>
<td>volumetric flow rate of RAE</td>
<td>L³ t⁻¹</td>
<td>74.2 L min⁻¹</td>
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<table>
<thead>
<tr>
<th>Subscripts</th>
<th>Description</th>
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<tr>
<td>$ox$</td>
<td>oxidized redox-active species</td>
</tr>
<tr>
<td>$red$</td>
<td>reduced redox-active species</td>
</tr>
<tr>
<td>$1$</td>
<td>solid (electrode) phase</td>
</tr>
<tr>
<td>$2$</td>
<td>liquid (electrolyte) phase</td>
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<tr>
<td>$high$</td>
<td>high-potential side</td>
</tr>
<tr>
<td>$low$</td>
<td>low-potential side</td>
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Table A1. Cont.

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
<th>Dimensions 1</th>
<th>Default Value 2</th>
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<tr>
<td>α</td>
<td>transfer coefficient</td>
<td>L^2 t^{-1}</td>
<td>0.5</td>
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<tr>
<td>ν</td>
<td>kinematic viscosity</td>
<td>m^2 s^{-1}</td>
<td>0.5</td>
</tr>
<tr>
<td>π_1</td>
<td>mass transfer correlation parameter</td>
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<td></td>
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<tr>
<td>κ RAET</td>
<td>RAE conductivity</td>
<td>T^3 M^{-1} L^{-3}</td>
<td>270 mS cm^{-1} 8</td>
</tr>
<tr>
<td>κ_{eff}</td>
<td>effective RAE conductivity</td>
<td>T^3 M^{-1} L^{-3}</td>
<td>5 m^{-1}</td>
</tr>
<tr>
<td>κ_{mem}</td>
<td>membrane conductivity</td>
<td>M^{-1} L^{-3}</td>
<td>67 mS cm^{-1} 110</td>
</tr>
<tr>
<td>µ</td>
<td>dynamic viscosity</td>
<td>kg m^{-1} s^{-1}</td>
<td>5 mPa s 8</td>
</tr>
<tr>
<td>η_{pump}</td>
<td>pump efficiency</td>
<td></td>
<td>70%</td>
</tr>
<tr>
<td>Δt</td>
<td>time step for cycling</td>
<td>t</td>
<td>20 s</td>
</tr>
<tr>
<td>ε</td>
<td>electrode porosity</td>
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<td>85% 73</td>
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<tr>
<td>Δφ_m</td>
<td>overpotential at the membrane</td>
<td>M L^{-3} t^{-3}</td>
<td>V</td>
</tr>
<tr>
<td>ΔP</td>
<td>pressure drop</td>
<td>M t^{-2} L^{-1}</td>
<td>Pa</td>
</tr>
<tr>
<td>ρ</td>
<td>RAE density</td>
<td>M L^{-3}</td>
<td>1.5 g mL^{-1} 8</td>
</tr>
<tr>
<td>σ</td>
<td>electrode conductivity</td>
<td>T^2 M^{-1} L^{-3}</td>
<td>5 m^{-1}</td>
</tr>
<tr>
<td>φ_i</td>
<td>potential in phase i</td>
<td>M L^2 I^{-1} t^{-3}</td>
<td>V</td>
</tr>
</tbody>
</table>

1 Dimension key: mass (M), length (L), time (t), temperature (T), current (I), matter (N). 2 In the case of a time-variant variable, the initial value is given. In the case of multiple values being assumed or estimated based on other system parameters, only the default SI units are given. 3 Estimated based on long thin-fiber characteristics based on $n_e = 4 \times (1 - \epsilon)/d_f [73,91]$. 4 This flow rate was selected by scaling the flow rate from a few references [58,62], as well as our own experimental practices. Heuristically, it appears that a flow rate of ~2 milliliters per minute per square centimeter of cell area is fairly standard practice. 5 Estimated using a Bruggeman relation $\kappa_{eff} = k''^{1.5} [72,73,90]$.

Figure A2. Clarifying illustration of system geometry parameters that enable extension beyond single-cell analysis.

Figure A1. Clarifying illustration of system geometry parameters that enable extension beyond single-cell analysis.
References


28. Viswanathan, V.; Crawford, A.; Maupin, G.; Sprenkle, V. 1 kW/1 kWh advanced vanadium redox flow battery utilizing mixed acid electrolytes.


73. Barton, J.L.; Milbsite, J.D.; Hinricher, J.J.; Brushett, F.R. Quantifying the impact of viscosity on mass-transfer coefficients in redox flow batteries. *J. Power Sources* **2018**, 399, 133–143. [CrossRef]


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