A grid-free approach for simulating sweep and cyclic voltammetry

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ABSTRACT
We present a computational approach to simulate linear sweep and cyclic voltammetry experiments that does not require a discretized grid in space to quantify diffusion. By using a Green’s function solution coupled to a standard implicit ordinary differential equation solver, we are able to simulate current and redox species concentrations using only a small grid in time. As a result, where benchmarking is possible, we find that the current method is faster than (and quantitatively identical to) established techniques. The present algorithm should help open the door for studying adsorption effects in inner sphere electrochemistry.

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I. INTRODUCTION
A. Models for outer-sphere electrochemical electron transfer (ET) and voltammetry

A common experimental technique throughout the field of electrochemistry is sweep voltammetry, where the electrostatic potential of a metal electrode (relative to an electrochemical cell containing an electroactive redox species) is changed with time. The resultant current is usually plotted as a function of the change in driving force. When the potential is ramped in a linear fashion, \( \mathcal{E}(t) = \mathcal{E}_0 + \nu t \), where \( \nu \) is the scan rate, the resulting plot is known as a linear sweep voltammetric (LSV) curve. When the scan is carried out in both directions (oxidation and reduction), it is known as cyclic voltammetry (CV). The current in such experiments arises from both (i) the change in the electron transfer (ET) kinetics at the electrode (due to the change in the driving force for the ET reaction) and (ii) the diffusional effects that create a concentration gradient of the redox active species in solution. This combination of effects driving the current vs potential profile makes linear sweep voltammetry an ideal experiment for inferring unknown chemical and physical parameters of interest in the electrochemical system. In particular, nowadays, it is routine to use LSV to infer the diffusion constant of the redox active species\(^2\) or the standard rate constant for an ET reaction, \( k_0 \) (the rate constant where the forward and backward rates for the redox reactions are equal).\(^3\)

In order to properly analyze an LSV experiment, one requires a mathematical model with a reasonable description of both the diffusional and ET effects. Consider the most basic electrochemical reaction, the reduction of a redox species \( A \) to its reduced form \( B \),

\[
A + e^{-} \underset{k_0}{\overset{k_f}{\rightleftharpoons}} B,
\]

where species \( A \) exists in bulk solution but, before a reaction, must be brought to the metal surface. For most simulations, diffusional effects are modeled through a standard parabolic, second order partial differential equation (PDE),\(^4\)

\[
\frac{\partial c_A}{\partial t} = D_A \frac{\partial^2 c_A}{\partial x^2},
\]

\[
\frac{\partial c_B}{\partial t} = D_B \frac{\partial^2 c_B}{\partial x^2}.
\]

Thereafter [see Eq. (8)], ET effects are incorporated through either a source term in the PDE or electrode boundary conditions. This approach has widespread applicability and can be adapted to take into account electrical double layer (EDL) effects (which are also called Frumkin corrections to the rate constant, which account for the uncompensated resistance due to the field drop at the point/plane of ET),\(^5\) surface adsorption of a species pre-ET,\(^6\)\(^7\) coupled homogenous chemical reactions,\(^8\)\(^9\) and many...
other effects as well. Some of the relevant PDEs can be solved analytically if the ET kinetics are considered to be infinitely fast (i.e., when ET is taken to be instantaneous and dependent upon only the thermodynamics of the redox couple at a given external potential); this is known as the Nernstian limit (again, see below). Realistically, however, most ET kinetics are not infinitely fast, and the Nernstian limit is an idealization. In practice, in order to interpret his/her data, the experimentalist will usually need to model voltammetry experiments explicitly, i.e., one requires a computer to solve the necessary PDEs [e.g., in Eq. (8)] that incorporate (at minimum) both diffusion and a time-dependent ET rate constant.

Finally, for completeness, a few words are appropriate as far as how most electrochemical simulations model electron transfer if one wishes to go beyond the Nernstian limit. In general, there are two common choices for the rate constant: Butler–Volmer (BV) and Marcus–Hush–Chidsey (MHC).15

1. One common choice for the ET rate constant is the Butler–Volmer (BV) expression. BV dynamics are characterized by two parameters: the charge transfer coefficient, $\alpha$, and the standard prefactor $k^0$. The charge transfer coefficient’s physical interpretation has historically been a matter of some debate, but, mathematically, it is defined as the change in the log of the current with respect to the change in the driving force (in unitless form): $\alpha = \left[ \frac{dI}{dF} \right]$. The total BV forward ($k_f$) and backward ($k_b$) rate constants take the form

$$
k_f = k^0 e^{-\alpha F (E^f - E'^f)},$$
$$k_b = k^0 e^{\alpha F (E^f - E'^f)},$$

where $E^f$ is the formal potential (which is dependent on the experimental setup), $F$ is the Faraday constant, and $R$ is the universal gas constant.

2. The other common form for the ET rate constant is the MHC expression. Like BV, the MHC expression can also be characterized by two parameters, the prefactor $k^0$ and the reorganization energy $\lambda$. While the prefactor $k^0$ has the same interpretation as in BV theory, the reorganization energy $\lambda$ is defined as the free energy change associated with relaxing the local solvent environment from one configuration (equilibrated to the reactant state) to another configuration (equilibrated to the product state) all the while keeping the molecule’s electronic state fixed as the product (see Fig. 1). Historically, BV theory has been the first choice among electrochemists due in large part to its inherent simplicity and ease of use. However, MHC theory has now become relatively common as the model has close ties to microscopic phenomena and many situations exist where BV theory is incapable of capturing the correct ET behavior.
   - For homogenous ET in solution, at high temperatures, MHC theory is roughly equivalent to transition state theory (insofar as they are exponential in the barrier height) such that

$$k = k^0 e^{-\frac{\lambda}{R T}},$$

where $\Delta G^\ddagger$ is the barrier height associated with the transition state, while $k_0$ and $T$ have their usual definitions as the Boltzmann constant and temperature, respectively. While the exact form of $k^0$ can be a complicated integral expression and depend on the nature of the reaction (inner or outer sphere), such a distinction will not be relevant for our present purposes. There are multiple analytical series expansions that can be used to calculate the MHC rate constants,18 but these complicated scenarios can often be avoided by assuming symmetric, quadratic free energy surfaces. In practice, if one makes a parabolic free energy assumption, the Marcus rate takes on the famous form

$$k = k^0 e^{-\frac{(\Delta G)^2}{2\lambda}},$$

where $\Delta G$ is the free energy difference between the electron being located on the acceptor and that on the donor.
   - For heterogeneous electrochemical ET, the Marcus rate constant must be modified to account for the fact that the product is no longer a chemical species in solution but rather a metal electrode with a continuum of electronic states, which obeys an occupancy dictated by a Fermi distribution at a given external potential. By integrating the rate constants over the distribution of electronic states in the electrode, one obtains the so-called MHC rate constant for electrochemical ET.19
The equations given above are typically solved on a discretized one-dimensional grid. In order to conceptually separate nonlinear adsorption from the actual ET event, electrochemists have developed a few different model isotherms that quantify what concentration of a species is expected to be adsorbed on a given surface (Γi). The Langmuir isotherm, which assumes that at high enough solution concentrations, there is a saturating coverage of the electrode (denoted as Γs). The Langmuir isotherm further assumes that there are no interactions between the species on the electrode surface and that there is no surface heterogeneity. Mathematically, for a single species A, the Langmuir isotherm is given by

$$\frac{\Gamma_A}{\Gamma_s - \Gamma_A} = \beta_{A \text{ ads}} \iff \frac{\Gamma_A}{\Gamma_s} = \frac{\beta_{A \text{ ads}}}{1 + \beta_{A \text{ ads}}}, \quad (6)$$

where $\beta_A$ is the equilibrium constant for the adsorption/desorption of A, $\beta_A = \frac{k^A}{k^{A \text{ des}}}$. For the case of multiple species (for example, A and B, where B is the reduced form of A), the surface concentration of species $i$ ($i = A, B$) can be expressed as

$$\Gamma_i = \frac{\Gamma_A \beta_{A \text{ ads}}}{1 + \beta_{A \text{ ads}} + \sum_{j \neq i} \beta_{A \text{ des} j}}. \quad (7)$$

More complicated isotherms are sometimes used in the literature, such as the Frumkin isotherm (which allows for interactions between adsorbed species and the surface that can be coverage/concentration dependent). Below, we will present data from only the Langmuir isotherm (though the method itself will be quite general).

II. ALGORITHM

A. Discretized method

Before presenting our grid-free approach, we quickly review the standard approach for electrochemical voltammetry in the absence of adsorption. Assuming that all electron transfer occurs only at the boundary, the differential equation for the time evolution of the concentration of each species is

$$\frac{\partial c_i}{\partial t} = \frac{D_A}{\partial x^2} c_i - (k_f c_A - k_d c_B) \delta(x), \quad (8)$$

where $c_A$ and $c_B$ are the normalized concentrations of species $A$ and $B$, respectively, and $D_A$ and $D_B$ are the diffusion constants for species $A$ and $B$. The delta function source term is used to enforce that ET occurs only at the solution–electrode interface, defined as $x = 0$. The equations given above are typically solved on a discretized one-dimensional grid. In other words, for a discrete set of points $[x_1, x_2, \ldots, x_N]$ with spacing $\Delta x$ and a time step $\Delta t$, one solves for $c_i(\Delta x, \Delta t)$ in a big loop. In principle, it is usually trivial to solve the differential equation in Eq. (8) using the Euler method with a small enough time step. Unfortunately, however, the ET source terms at the metal electrode [i.e., those terms proportional to $\delta(x)$] usually lead to a myriad of numerical problems for any realistic time discretization (that would certainly dissuade most undergraduates). Moreover, we cannot emphasize enough that electrochemistry occurs over a vast array of time scales. While electron transfer may occur on the time scale of picoseconds, an electrochemical sweep over 1 V with a scan rate of 10 mV/s occurs over 100 s; one must propagate Eq. (8) over an immensely long time range.

The most common solution to such numerical difficulties is to transform the source terms in the above equation into boundary conditions, utilizing a no-flux boundary condition at the electrode surface; for details, see Ref. 22. Thereafter, one must still invoke an implicit ordinary differential equation (ODE) solver as the equations are stiff. Alternatively, because the operator on the right-hand side of Eq. (8) is linear, the simplest approach is to just diagonalize the corresponding (non-Hermitian) operator for diffusion and ET. While diagonalization of a non-Hermitian operator is usually unstable, it is well known that the non-Hermitian diffusion operator (in one dimension) can be transformed into a Hermitian operator. As a result, if one is prepared to work with a grid of points in one spatial dimension and one is not concerned with adsorption, there are a few numerically stable techniques available as far calculating LSV curves. Our recent work has shown that one can use such a brute force approach to capture more complicated electrochemical scenarios, including proton coupled electron transfer.

B. Grid-free method

Unfortunately, the methods above face great difficulties when nonlinear adsorption and desorption are considered. On the one hand, the presence of non-linear Langmuir isotherms leads to an even stronger stiffening of the coupled set of PDEs such that using an implicit differential equation solver to solve for the dynamics on a grid often fails. On the other hand, diagonalization is obviously impossible with nonlinear equations of motion. As a result,
modeling adsorption is daunting for most parameter sets. The most successful algorithm to date comes from Yang and Compton, who have used a spatial grid in a recent article to study phase transitions of adsorbed electroactive species.\textsuperscript{10}

The goal of the present letter is to present a different solution to this problem, one that completely removes the necessity for a grid in space. In so doing, we will vastly reduce the number of variables that must be propagated in time; rather than spending time solving the heat equation in solution (and calculating the surface concentration \(\Gamma_b\)) we will focus our attention exclusively on the surface where the electron transfer occurs (and calculate the surface concentration \(\Gamma_A\)).

1. The relevant PDEs

Before outlining our new approach, let us define the differential equations for coupling diffusion, ET, and adsorption that we wish to solve. The PDEs for the solution concentrations \(c_A\) and \(c_B\) take the standard form \([\text{following Eq. (8)}]\), with the addition of adsorption/desorption rates, \(v_{\text{ads}}/v_{\text{des}}\) at the electrode surface \(x = 0\),

\[
\frac{\partial c_A}{\partial t} = D_A \frac{\partial^2 c_A}{\partial x^2} - \left( k_{\text{ads}}^A (t) c_A - k_{\text{des}}^A (t) c_B \right) + \left( v_{\text{ads},A} - v_{\text{des},A} \right) \delta(x),
\]

(9a)

\[
\frac{\partial c_B}{\partial t} = D_B \frac{\partial^2 c_B}{\partial x^2} + \left( k_{\text{ads}}^B (t) c_A - k_{\text{des}}^B (t) c_B \right) - \left( v_{\text{ads},B} - v_{\text{des},B} \right) \delta(x).
\]

(9b)

The ODEs for the surface concentrations, \(\Gamma_A\) and \(\Gamma_B\), involve both electron transfer and surface adsorption/desorption,

\[
\frac{\partial \Gamma_A}{\partial t} = v_{\text{ads},A}(t) - v_{\text{des},A}(t) - \left( k_{\text{ads}}^A (t) \Gamma_A - k_{\text{des}}^A (t) \Gamma_B \right).
\]

(10a)

\[
\frac{\partial \Gamma_B}{\partial t} = v_{\text{ads},B}(t) - v_{\text{des},B}(t) + \left( k_{\text{ads}}^B (t) \Gamma_A - k_{\text{des}}^B (t) \Gamma_B \right).
\]

(10b)

The choice of isotherm will dictate the exact form of \(v_{\text{ads,i}}/v_{\text{des,i}}\); e.g., for the case of a Langmuir isotherm, we choose

\[
v_{\text{ads},A}(t) = k_{\text{ads}}^A c_A(t, x = 0) \left[ \Gamma_A - (\Gamma_A + \Gamma_B(t)) \right],
\]

(11a)

\[
v_{\text{des},A}(t) = k_{\text{des}}^A \Gamma_A(t),
\]

(11b)

\[
v_{\text{ads},B}(t) = k_{\text{ads}}^B c_B(t, x = 0) \left[ \Gamma_A - (\Gamma_A + \Gamma_B(t)) \right],
\]

(11c)

\[
v_{\text{des},B}(t) = k_{\text{des}}^B \Gamma_B(t).
\]

(11d)

\(k_{\text{ads}}^A/k_{\text{des}}^A (k_{\text{ads}}^B/k_{\text{des}}^B)\) can follow different ET rate laws and are parameterized in different ways (different \(a/\Lambda, R^0, E^0\), etc.) to reflect the correct underlying physics.

Finally, when calculating the current, note that, in principle, there will be separate contributions from both solution and adsorbed ET,

\[
I(t) = nFA \left[ \left( k_{\text{ads}}^A (t) c_A(t, x = 0) - k_{\text{des}}^A (t) c_B(t, x = 0) \right) + \left( k_{\text{ads}}^B (t) \Gamma_A(t) - k_{\text{des}}^B (t) \Gamma_B(t) \right) \right],
\]

(12)

where \(n\) is the number of electrons, \(A\) is the surface area of the electrode, and \(F\) is the Faraday constant. The time dependence of the rate constants \(k_f\) and \(k_b\) enters only through their dependencies on the overpotential (which is linearly changed as a function of time in a sweep voltammetry experiment). Formally, the current depends only on four variables at a single time: \(c_A(x = 0, t), c_B(x = 0, t), \Gamma_A(t), \) and \(\Gamma_B(t).\)

2. A Green’s function approach

Given that \(v_{\text{ads,i}}/v_{\text{des,i}}\) and \(I(t)\) depend only on \(c_A(x = 0), c_B(x = 0), \Gamma_A, \) and \(\Gamma_B\) [and specifically not on any of the remaining \(c_A(x \neq 0)\) and \(c_B(x \neq 0)\)], our goal is to find a maximally efficient algorithm that tracks only these four values (without tracking the remaining solution concentrations away from the electrode surface). Here, for simplicity, we will assume that species \(A\) and \(B\) have equal diffusivities \(D_A = D_B\); all of the following results can be simply modified to accommodate unequal diffusion constants if needed.

Following the standard practice in modeling diffusion,\textsuperscript{4} we will introduce a Green’s function formalism in order to integrate Eqs. (9)–(11) without a spatial grid. Note that in the absence of adsorption, Green’s functions and Laplace transforms have been used previously by Bazant \textit{et al.} to model CV curves for the case of thin double sweep voltammetry. In particular, various analytical CV results can be obtained in the Nernstian limit,\textsuperscript{16} especially if one focuses on processes with one ion of interest. Here, we will extend such approaches to the case of several ions with adsorption where no analytic result is possible. We begin by invoking Duhamel’s principle\textsuperscript{27} to recast our coupled differential equations modeling \(c_A\) and \(c_B\) as integral equations. For an inhomogeneous diffusion equation of the form expressed in Eq. (9), with initial and boundary conditions \(c_B(t = 0, x) = c_{B0}(t = 0, x) = 0\) and \(c_A(t = 0, x) = c_{A0}(t = x = 0) = 1\), one can formally invert these equations of motion to solve for the solution concentration in integral form (we will show \(c_B\), but the derivation is identical for \(c_A\)).

\[
c_B(x,t) = \int_0^t \int_0^x G(x-y,t-s) \left[ \left( k_{\text{ads}}^B (s) c_B(s,y) - k_{\text{des}}^B (s) c_B(s,y) \right) - v_{\text{ads},B} + v_{\text{des},B} \right] \delta(y)ds dy,
\]

(13)

where

\[
G(x-y,t-s) = \frac{1}{\sqrt{4D_B \pi (t-s)}} e^{-\frac{(x-y)^2}{4D_B(t-s)}},
\]

(14)

is one half of the heat equation Green’s function in one dimension.\textsuperscript{27} Note that a factor of 2 appears here because the original diffusion equation spans only the half-space \(x \in [0, \infty)\).
In Eq. (13), the expressions for \( b(s, y) \) and \( a(s) \) are obtained directly from reorganization of Eq. (9); for a Langmuir isotherm, these expressions are
\[
\begin{align*}
 b(s, y) &= k^A_y(s_0) c_A(s_0) + k^B_y s_B(s), \\
 a(s) &= -(k^A_y(s_0) + k^B_y(s_0)) \Gamma_A(s_0) - (\Gamma_A(s) + \Gamma_B(s)).
\end{align*}
\]

(15)

Since we are only concerned with the solution concentration at \( x = 0 \), the equation we wish to numerically solve is given by
\[
c_a(t, 0) = \int_0^t \frac{1}{\sqrt{D_B \pi(t-s)}} [b(s, 0) + a(s)c_B(s, 0)] ds.
\]

(16)

We face two challenges in employing Eq. (16) to measure electrochemical currents:

1. \( b(s, 0) \) and \( a(s) \) depend on \( \Gamma_A \), \( \Gamma_B \), and \( c_A \) explicitly, while \( \Gamma_A \) and \( \Gamma_B \) depend explicitly on \( c_A \) and \( c_B \). See Eqs. (10) and (11).
2. There is a singularity in Eq. (16) at \( s = t \), which must be carefully accounted for due to the importance of accurate integral evaluation at this point.

Let us begin with the second challenge. To treat the singularity, we cast the integrand in Eq. (16) into a general form of \( f(s) = b(s, 0) + a(s)c_B(s, 0) \). We then break this term into two pieces: \( \frac{b(s, 0) - b(t)}{\sqrt{\pi(t-s)}} \) and \( \frac{b(t)}{\sqrt{\pi(t-s)}} \). The first piece goes to zero at \( s = t \), and the second piece can be integrated analytically, removing the singularity. By integrating analytically, the second piece reduces to \( \left[ f(t) \sqrt{t} - f(s) \sqrt{s} \right] = -2f(t) \sqrt{t} \). By using nonweighted numerical quadrature (according to the trapezoid rule), one can obtain a numerical prescription for calculating \( c_B(t) \) as a function of \( b(s, 0), a(s) \), and the prior history of \( c_B \) (expressed on a grid \( s_1, s_2, \ldots, s_N \), with spacing \( \Delta s \) and \( s_N = t \)),
\[
 c_B(t, 0) = \frac{\Delta s \sum_{i=1}^{N-1} \frac{f(s_i) - f(s_{i+1})}{2\sqrt{s_i}}}{\sum_{i=1}^{N-1} \frac{1}{\sqrt{s_i}}}.
\]

(17)

For the derivation, see the Appendix. Note that in the absence of adsorption, Eq. (17) is entirely sufficient for calculating the current as a function of time/voltage for a sweep voltammetry simulation—without any need for solving an ODE on a discretized grid in space. For instance, suppose that \( A \) and \( B \) have equal diffusion constants for \( A \) and \( B \). In such a case, \( c_A(t, 0) + c_B(t, 0) = c_A(x = \infty) + c_B(x = \infty) = c_{bulk} \); thus, \( c_A(t, x = 0) \) can be determined by \( c_A(t, x = 0) = c_{bulk} - c_B(t, 0) \). Therefore, since \( k^A_y \) and \( k^B_y \) are analytic expressions...
TABLE I. Computational wall time obtained by the method (Python).

<table>
<thead>
<tr>
<th>Scan rate (V/s)</th>
<th>No adsorption (s)</th>
<th>Adsorption (s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>0.05</td>
<td>3.70</td>
<td>22.06</td>
</tr>
<tr>
<td>0.1</td>
<td>4.89</td>
<td>21.13</td>
</tr>
<tr>
<td>1</td>
<td>3.40</td>
<td>20.54</td>
</tr>
<tr>
<td>5</td>
<td>3.67</td>
<td>19.87</td>
</tr>
</tbody>
</table>

TABLE II. Computation time obtained by the method (Python).

<table>
<thead>
<tr>
<th>Algorithm</th>
<th>Wall time (s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Grid-free</td>
<td>2.34</td>
</tr>
<tr>
<td>Discretized, $N_x = 50$</td>
<td>9.94</td>
</tr>
<tr>
<td>Discretized, $N_x = 100$</td>
<td>30.91</td>
</tr>
<tr>
<td>Discretized, $N_x = 200$</td>
<td>83.02</td>
</tr>
</tbody>
</table>

already needed for evaluation of Eq. (17), one can easily calculate the current at time $t$ using Eqs. (12) and (17), and thus, challenge No. 1 disappears. Implementing Eq. (17) is as fast as (or faster than) any implicit ODE based scheme for evaluating equations of the type in Eq. (8) and requires no black-box type assumptions that are often inherent when implicitly solving stiff ODEs with common numerical software packages.\footnote{Note that Eq. (17) can be written as $c_B(t,0) = \frac{\delta}{\Gamma A(t) + \Gamma B(t)}$ and $c_B(t,0)$ at a given time step until the end of the simulation is reached.}

Next, let us consider the first challenge, i.e., how to self-consistently calculate concentrations when there is adsorption of redox active species. In order to simultaneously solve Eq. (10) (a standard ODE) and Eq. (17) (an integro-differential equation) in a coupled fashion, we will apply the following procedure:

1. At $t = dt$, calculate $c_B(dt,0)$ using Eq. (17) and the values of $\Gamma_A$, $\Gamma_B$, and $c_A$ at their prior time step values.
2. Calculate $c_A(dt,0)$ using the analogous equation (17) for $c_A$.

3. Calculate $\Gamma_A(dt)$ and $\Gamma_B(dt)$ using Eq. (10) and an implicit ODE solver (in this case, a backward differentiation formula of order 5).

4. Return to step 1 of this procedure until simultaneous convergence to a predetermined tolerance is achieved for all four values $c_A(dt,0)$, $c_B(dt,0)$, $\Gamma_A(dt)$, and $\Gamma_B(dt)$ within one single convergence step.

5. Move on to the next time step, repeating steps 1–5 at each time step until the end of the simulation is reached.

Note that, in step (1), for the first guess of $\Gamma_A$ and $\Gamma_B$ at a given time step, extrapolation methods can be used to obtain a better initial guess rather than simply using the prior time step; this usually yields much faster convergence. In our experience, linear extrapolation using two prior times $[\Gamma_A(t) = 2\Gamma_A(t - \delta t) - \Gamma_A(t - 2\delta t)]$ yields a substantial improvement in convergence time, while higher order extrapolations contribute little for accelerating convergence.

In Fig. 2, we show a set of illustrative examples of CVs generated with this methodology, both in the case (a) without and (b) with adsorptive effects. From the voltammograms in Fig. 2(a), without adsorption, at low scan rates, we see the expected peak separation of 57 mV for a reversible ET reaction in solution, and the peak separation increases slightly as the scan rate increases and approaches the irreversible ET regime. Conversely, in Fig. 2(b), we see no peak separation in the case where ET occurs between surface bound species, as expected.\footnote{These results demonstrate that the above algorithm can capture the behavior of both coupled mass and electron transfer in solution and at surface bound electrochemical interfaces under a sweeping potential. Unfortunately, these simulations should also make clear that any attempt to invert\footnote{Similar to the above argument about the scan rate for traditional grid-based methods, more spatial points imply larger computational cost, whereas for the method outlined in this paper, no such discretization is required as there is no added computational cost. All simulations were performed in Python, using the NumPy and SciPy libraries, on a 2.8 GHz Dual-Core Intel Core i5 processor with 16 GB of RAM.} a voltammogram to back out the relevant parameters in Eqs. (9)–(11) would be extremely difficult.}

Additionally, to demonstrate the computational benefit of the present algorithm, below we show two tables illustrating the computational improvement of this method. Table I shows computation times for the results shown in Fig. 2, where we see that the present algorithm is relatively invariant to the scan rate—even with adsorption. This behavior stands in contrast to the typical grid-based ODE methods, which can require longer times at smaller scan rates, as more time is needed to propagate dynamics using an implicit ODE solver in order to cover a given potential range. Finally, Table II shows the relative improvement in computation time for the present algorithm as compared to the more traditional approach where one solves the spatial grid-based coupled ODEs through discretization of the relevant propagator over a set of grid points and diagonalization of the diffusion operator.\footnote{Table II. Computation time obtained by the method (Python).}

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</table>

III. CONCLUSION

We have introduced an integral based approach for simulating electrochemical ET to model ET along with adsorption of redox species. Our approach is in quantitative agreement with grid-based PDE solutions for a single ET experiment; more generally, however, the present algorithm should allow us to simulate more complicated sweep experiments that involve adsorption kinetics and beyond. Looking forward, one can imagine several applications of the present methodology. First, one would like to study the various parameter regimes of electrochemical voltammetry experiments where adsorption plays a key role. After all, the addition of another time scale/parameter set corresponding to adsorption/desorption rates, plus the potential asymmetry in the redox behavior of charge/uncharged species on a surface, implies that a realistic description of electrochemical dynamics and kinetics can and must go beyond the standard Nernstian and irreversible regimes that we have all grown up with. Second, one can imagine

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extending the present methodology to two and three dimensional surfaces, investigating how geometry and defects in the electrode surface affect current; indeed, previous work with diagonalization has precluded such large scale simulation, but Green’s function techniques in three dimensions have proven successful as far as modeling heat transfer in environments with finite domains.\(^7\) Third and finally, integrating the present diffusion equations with microscopic (atomistic) models of interfaces remains very difficult, especially if one wishes to construct \textit{ab initio} models of bond-making and bond-breaking electrochemistry, and for maximum impact, the present algorithm will certainly need to be extended to far larger simulations with more intermediates and even more dynamical time scales.

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**APPENDIX: DERIVATION OF EQ. (17)**

Below, we will carry out a general derivation of Eq. (17) from Eq. (16), for either species A or B, with arbitrary initial conditions. We first break the integrand into two pieces, \(\frac{f(t)-f(t)}{\sqrt{t-t}}\) and \(\frac{f(t)}{\sqrt{t-t}}\), where \(f(s) = b(s, 0) + a(s)\Delta c(s, 0)\). The first integral is evaluated using the trapezoidal rule, while the second integral is evaluated analytically, 

\[
\frac{\Delta}{(\pi D)^{3/2}} \int_0^t \left[ \frac{b(s, 0) + a(s)\Delta c(s, 0) - b(t, 0) + a(t)\Delta c(t, 0)}{\sqrt{t-s}} \right] ds + \frac{b(0, 0) - b(t, 0) + a(t)\Delta c(t, 0)}{2\sqrt{t}} + 2\sqrt{t}(b(t, 0) + a(t)\Delta c(t, 0)).
\]

Rearrangement of all \(c_j(t, 0)\) terms to the left-hand side yields

\[
c_j(t, 0) = c_j(0, 0) + \frac{(D_j \pi)^{3/2}}{\sqrt{t}} + \Delta \left[ \sum_{i=1}^{N-1} \frac{a(t)}{\sqrt{t-s_i}} + \frac{a(t)}{2\sqrt{t}} \right] - c_j(0, 0) \pi \left[ \sum_{i=1}^{N-1} \frac{b(s, 0) + a(s)\Delta c(s, 0) - b(t, 0)}{\sqrt{t-s_i}} + \frac{0.5(b(0, 0) - b(t, 0) + a(t)\Delta c(t, 0))}{\sqrt{t}} \right] + 2b(t, 0)\sqrt{t}.
\]

Finally, the isolation of \(c_j(t, 0)\) yields the final result

\[
c_j(t, 0) = c_j(0, 0) + \frac{(D_j \pi)^{3/2}}{\sqrt{t}} + \Delta \left[ \sum_{i=1}^{N-1} \frac{a(t)}{\sqrt{t-s_i}} + \frac{a(t)}{2\sqrt{t}} \right] - c_j(0, 0) \pi \left[ \sum_{i=1}^{N-1} \frac{b(s, 0) + a(s)\Delta c(s, 0) - b(t, 0)}{\sqrt{t-s_i}} + \frac{0.5(b(0, 0) - b(t, 0) + a(t)\Delta c(t, 0))}{\sqrt{t}} \right] + 2b(t, 0)\sqrt{t}.
\]

Note that Eq. (17) is recovered exactly when the index \(j\) is replaced with \(B\) and the initial condition \(c_B(0, 0) = 0\) is substituted.

**DATA AVAILABILITY**

The data that support the findings of this study are available within the article. The code implementing this algorithm can be found at https://github.com/alecja/electrochemistry_sim.

**REFERENCES**


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When $D_A \neq D_B$, then one must use the more strict condition that there is no net flux $D_A \frac{dc_A}{dx} = -D_B \frac{dc_B}{dx}$, which implies $D_A c_A = D_B c_B$ is a constant in space.