Supplementary Information for "The Application of Physics-Informed Neural Networks to Hydrodynamic Voltammetry"

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1 PINN Simulation of a Double Channel Electrode

The steady state transport limited currents flowing through both the generator and detector electrodes and thus the collection efficiency of double channel electrode. $C_A(X,Y)$ is predicted at steady state using a PINN. The physical constraints are:

$$\begin{aligned} \frac{\partial^2 C_A}{\partial Y^2} + \frac{\partial^2 C_A}{\partial X^2} - V_X \frac{\partial C_A}{\partial X} &= 0, \, \Omega_X \times \Omega_Y \quad 1.1 \\ \frac{d C_A}{d Y} &= 0, \, X = [0, X_1], \, Y = 0 \quad 1.2 \\ C_A &= 0, \, X = [X_1, X_2] \quad 1.3 \\ \frac{d C_A}{d Y} &= 0, \, X = [X_2, X_3], \, Y = 0 \quad 1.4 \\ C_A &= 1, \, X = [X_3, X_4] \quad 1.5 \\ \frac{d C_A}{d Y} &= 0, \, X = [X_4, X_{Channel}], \, Y = 0 \quad 1.6 \\ C_A &= 1, \, X = 0, \Omega_Y \quad 1.7 \\ \frac{d C_A}{d Y} &= 0, \, \Omega_X, Y = Y_{channel} \quad 1.8 \end{aligned}$$

where $\Omega_X \in [0, X_{channel}]$ and $\Omega_Y \in [0, Y_{channel}]$. To enforce

the constraints, the mean square error (MSE) loss functions are:

$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{\partial^{2}C_{A}}{\partial Y^{2}} + \frac{\partial^{2}C_{A}}{\partial X^{2}} - V_{x}\frac{\partial C_{A}}{\partial X}\right)^{2}, \Omega_{X} \times \Omega_{Y} \quad 2.1$$
$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{dC_{A}}{dY}\right)^{2}, X = [0, X_{1}], Y = 0 \quad 2.2$$

$$\frac{1}{N}\sum_{i=1}^{N} (C_A)^2, X = [X_1, X_2]$$
 2.3

$$\frac{1}{N} \sum_{i=1}^{N} \left(\frac{dC_A}{dY}\right)^2, X = [X_2, X_3], Y = 0 \qquad 2.4$$

$$\frac{1}{N} \sum_{i=1}^{N} (C_A - 1)^2, X = [X_3, X_4]$$
 2.5

$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{dC_{A}}{dY}\right)^{2}, X = [X_{4}, X_{Channel}], Y = 0 \quad 2.6$$
$$\frac{1}{N}\sum_{i=1}^{N} (C_{A} - 1)^{2}, X = 0, \Omega_{Y} \quad 2.7$$
$$1\sum_{i=1}^{N} (dC_{A})^{2}, Z = 0, \Omega_{Y} \quad 2.7$$

$$\frac{1}{N} \sum_{i=1}^{N} \left(\frac{dC_A}{dY} \right)^2, \, \Omega_X, Y = Y_{channel}$$
 2.8

The global loss implemented in PINN is a linear combination of all the MSEs as:

$$L = \sum_{i=1}^{8} w_i MSE_i \#(3)$$
where w_i is the weight of each MSE. w is set to 1 to generate the results in this paper

paper.

PINN Simulation of a Channel Electrode with CE reaction 2

The transport limited current of channel electrode is calculated for a CE reaction via a PINN which predicts the concentration of the two species, $C_R(X,Y)$ and $C_A(X,Y)$ in the simulation domain while satisfying the following two sets of constraints. For species R:

$$\begin{split} \frac{\partial^2 C_A}{\partial Y^2} + \frac{\partial^2 C_A}{\partial X^2} - V_x \frac{\partial C_A}{\partial X} + K_f C_R - K_b C_A &= 0 \quad 4.1 \\ \frac{\partial C_A}{\partial Y} &= 0, X = [0, X_1], y = 0 \quad 4.1 \\ C_A &= 0, X = [X_1, X \boxtimes_2], y = 0 \quad 4.2 \\ \frac{\partial C_A}{\partial Y} &= 0, X = [X_2, X_{channel}] \quad 4.3 \\ C_A &= C_A^*, X = 0, Y = [0, Y_{channel}] \quad 4.4 \\ \frac{\partial C_A}{\partial Y} &= 0, X = [0, Y_{channel}], Y = Y_{channel} \quad 4.5 \\ \text{And for species } A: \end{split}$$

$$d_{R}\frac{\partial^{2}C_{R}}{\partial Y^{2}} + d_{R}\frac{\partial^{2}C_{R}}{\partial X^{2}} - V_{x}\frac{\partial C_{R}}{\partial X} - K_{f}C_{R} + K_{b}C_{A} = 0 \quad 5.1$$

$$\frac{\partial C_{R}}{\partial Y} = 0, X = [0, X_{channel}], Y = 0 \quad 5.2$$

$$C_{R} = C_{R}^{*}, X = 0, Y = [0, Y_{channel}] \quad 5.3$$

$$\frac{\partial C_{R}}{\partial Y} = 0, X = [0, X_{channel}], Y = Y_{channel} \quad 5.4$$

The MSE loss functions consider both eqn(4) and eqn(5) are:

$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{\partial^2 C_A}{\partial Y^2} + \frac{\partial^2 C_A}{\partial X^2} - V_X \frac{\partial C_A}{\partial X} + K_f C_R - K_b C_A \right)^2, \Omega_X \times \Omega_Y \qquad 6.1$$
$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{\partial C_A}{\partial Y} \right)^2, X = [0, X_1], y = 0 \qquad 6.2$$
$$\frac{1}{N}\sum_{i=1}^{N} (C_i)^2, X = [Y, X^{[2]}], y = 0 \qquad 6.3$$

$$\frac{1}{N} \sum_{i=1}^{N} (C_A)^2, X = [X_1, X \boxtimes_2], y = 0$$
6.3

$$\frac{1}{N} \sum_{i=1}^{N} \left(\frac{\partial C_A}{\partial Y} \right)^2, X = [X_2, X_{channel}]$$

$$6.4$$

$$\frac{1}{N} \sum_{\substack{i=1\\ \partial C_A}}^{N} (C_A - C_A^*)^2, X = 0, Y = [0, Y_{channel}]$$
6.5

$$\frac{\partial C_A}{\partial Y} = 0, X = [0, Y_{channel}], Y = Y_{channel}$$
6.6

$$\frac{1}{N}\sum_{i=1}^{N} \left(d_R \frac{\partial^2 C_R}{\partial Y^2} + d_R \frac{\partial^2 C_R}{\partial X^2} - V_X \frac{\partial C_R}{\partial X} - K_f C_R + K_b C_A \right)^2, \Omega_X \times \Omega_Y \quad 6.7$$
$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{\partial C_R}{\partial Y} \right)^2, X = [0, X_{channel}], Y = 0 \quad 6.8$$

$$\frac{1}{N}\sum_{i=1}^{N} (C_R - C_R^*)^2, X = 0, Y = [0, Y_{channel}]$$
6.9

$$\frac{1}{N}\sum_{i=1}^{N} \left(\frac{\partial C_R}{\partial Y}\right)^2, X = [0, X_{channel}], Y = Y_{channel}$$
6.10

The global loss is thus a linear combination of eqn(6) as:

$$L = \sum_{i=1}^{10} w_i MSE_i \, \#(7)$$

To predict mass transport of two species, two independent neural networks are first constructed, specializing in predicting C_R and C_A respectively. Then the two networks are connected via the physical constraints implemented according to eqn(6) and a schematic illustration is shown in Figure S 1. Implementation using TensorFlow can be found at <u>https://github.com/nmerovingian/PINN-Hydrodynamic-Voltammetry</u>.



Figure S 1. Schematic illustration of the structure of the PINN when predicting channel electrode transport limited current for a CE reaction. It shows two fully-connected neural network, each specializing in predicting the concentration of one species, which are then connected together by the physical constraints.

3 Computational time

The simulations were run using a Nvidia V100 GPU and 12 allocated CPU cores using the Advanced Research Computing (ARC) facility at the University of Oxford. We also performed the simulations on a desktop without a GPU and equipped with E5-2640 v4 CPU* 2 (20 CPU cores in total) and 16 GB of RAM. The computational time of the three models mentioned in the paper is compared in Table 1.

Table 1, Computational time at a cluster and a desktop

	Cluster	Desktop
Single microband channel electrode	1-2 hours	4-6 hours
Double microband channel electrode	0.5-1 hours	2-3 hours
Single microband channel electrode with CE reaction	~ 3 hours	~ 7 hours