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A generalized lateral interactions function to fit voltammetric peaks of self-assembled monolayers

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ARTICLE INFO

From the generalized lateral interactions model and via a few mathematical approximations, a suitable regression function has been developed in order to fit voltammetric peak of redox responsive SAM for extracting the characteristic parameters. The efficiency of the generalized lateral interactions function was tested on simulated and raw data.

1. Introduction

Since their initial discovery in 1946 [1], self-assembled monolayers (SAMs) have become an ideal system for the theoretical study of interfacial phenomena, especially in the field of electrochemistry when they are immobilized on conductive substrates like gold [2, 3]. A major constraint to obtain a fine characterisation of electroactive SAMs is the dispersion of the current-voltage characteristics around the fully solved theoretical one [4]. Indeed, cyclic voltammograms (CVs) of SAMs are most often complex with sharper or broader voltammetric peaks than in the ideal case [5-8].

Some of these electrochemical behaviours can be explained by several theories/models, like the generalized lateral interactions model (GLI model) [7]. They allow to clarify deviations from ideality and define characteristic parameters of theoretical voltammograms like peak potential (E_p), peak intensity (I_p) and full width at half maximum (FWHM) in the case of infinite electron transfer.

However, using the GLI model to perform a global fit of experimental CVs is impossible, because the expression of I-V curves leads to a recursive function, excluding a curve fitting analysis.

A first alternative to analyse CV peaks consists in using the graphical analysis tools of the commercial instrument control software. This method allows determining characteristic parameters (E_p, I_p and FWHM) of the CVs but do not allow determining if a particular experimental profile follows a theoretical curve.
A second alternative, detailed in previous work, proposed to fit CV peaks via usual or unusual functions [9, 10] such as Gaussian [11-13], Lorentzian [11-13], or Generalized Gaussian Function [14] but these calculations were either very complex or not based on a theoretical support.

Herein, we propose a regression function, derived from the GLI model [6], which allows extracting characteristic parameters of a voltammetric peak and interaction constants of an electroactive system by curve fitting.

2. The GLI model

In previous works, we presented theoretical studies to complete the lateral interaction model proposed by E. Laviron [5], by extending this initial model to non-random distributions of electroactive species and by taking into account the interactions between redox and non-redox species [7].

In the case of interactions between redox species only, this model enables current-voltage behaviours to be simulated and allows extracting characteristics parameters of CVs obtained from any surface distribution of electroactive SAM.

For a better reading of the article, we must summarize previous works.

The generalized lateral interactions model can be defined according to the main following hypotheses:

- The electroactive species are distributed on substrate with a unimodal statistical distribution of electroactive neighbours. A parameter \( \phi(\theta) \), between 0 and 1, defined for a normalized surface coverage \( \theta = \frac{\Gamma}{\Gamma_{\text{max}}} \), quantifies the segregation level of the electroactive species. For a randomly distributed SAM, \( \phi(\theta) = \theta \), and when a segregation exists on the surface, \( \phi(\theta) > \theta \),

- The sum of normalized surface coverage \( \theta_O \) and \( \theta_R \) of oxidized (O) and reduced (R) species is constant and equal to \( \theta \),

- The surface occupied by one molecule of O is equal to the surface occupied by one molecule of R,

- The electrochemical rate constant \( k_\pm \) is independent of the coverage,

- \( a_{OO}, a_{RR} \) and \( a_{OR} \) are the interaction constants between molecules of O, molecules of R and molecules of O and R, respectively. \( a_{ij} \) is positive for an attraction and negative for a repulsion.

For a full reversible reaction \( \left( \frac{k_+}{v} \to \infty \right) \), CVs are perfectly reversible and the current can be expressed as a function of \( \theta_O \):
The characteristic parameters of CVs as full width at half maximum (FWHM), peak potential \( (E_p) \) and peak current \( (i_p) \) can be extracted and defined as:

\[
E_p(\phi(\theta)) = E_o + \frac{RT}{nF} S \phi(\theta)
\]  \( (2) \)

\[
\text{FWHM}(\phi(\theta)) = \frac{2 \text{RT}}{nF} \left[ \ln \left( \frac{1+ \frac{2-G\phi(\theta)}{4-G\phi(\theta)}}{1- \frac{2-G\phi(\theta)}{4-G\phi(\theta)}} \right) - G\phi(\theta) \right] \approx \frac{\text{RT}}{nF} \left( 2 \ln \left( 2 \sqrt{2} + 3 \right) - \frac{3 \sqrt{2}}{2} G\phi(\theta) \right)
\]  \( (3) \)

\[
i_p(\theta, \phi(\theta)) = \frac{n^2 F^2 v A \Gamma_{\text{max}}}{RT} \frac{\theta}{2(2-G\phi(\theta))}
\]  \( (4) \)

With \( n, F, v, A, R, T, E, E_o \) have their usual meanings

And \( G = a_{\infty} + a_{RR} - 2a_{OR} \) and \( S = a_{RR} - a_{\infty} \) with \( |G| \) and \( |S| \leq 2 \)

\( G \) and \( S \) are defined as "global interaction" parameters and play an important role when associated with \( \phi(\theta) \). Indeed, \( G \cdot \phi(\theta) \) defines the shape of the peak and modulates the FWHM and the peak intensity \( (i_p) \), whereas \( S \cdot \phi(\theta) \) only defines the position of the peak potential \( (E_p) \).

3. The generalized lateral interactions function (GLI function)

3.1. Introduction to the GLI function

The GLI model is very useful to predict characteristic parameters of CVs but, in order to fit CVs, and contrary to the expression 1, the current needs to be expressed as a function of the applied potential \( (E) \).

For a fast reversible system \( \left( \frac{k^v}{v} \rightarrow \infty \right) \) the Nernst equation is applicable using surface activity of the two redox species [6]:

\[
\frac{\gamma_o}{\gamma_r} = \exp \left[ \frac{nF}{RT} (E - E_o) \right] = \frac{\theta_o}{\theta - \theta_o} \exp \left[ -2 \frac{\phi(\theta)}{\theta} (G\theta_o + \theta(a_{OR} - a_{RR})) \right]
\]  \( (5) \)

This expression directly links the applied potential with the normalized surface coverage of oxidized species \( (\theta_o) \) and interaction parameters \( (G, a_{\infty} \text{ and } a_{OR}) \).

From the equations 2 and 5, it is possible to optimise this expression in order to limit the number of unknown parameters:
\[
\frac{nF}{RT}(E - E_p) = \ln \left( \frac{\theta_o}{\theta - \theta_o} \right) - 2G \frac{\phi(\theta)}{\theta} \theta_o + G\phi(\theta) 
\]  
(6)

\[
\frac{\theta_o}{\theta - \theta_o} = \exp \left( \frac{nF}{RT}(E - E_p) + 2G \frac{\phi(\theta)}{\theta} \theta_o - G\phi(\theta) \right) = \exp(X) 
\]  
(7)

with \( X = \frac{nF}{RT}(E - E_p) + 2G \frac{\phi(\theta)}{\theta} \theta_o - G\phi(\theta) \)

\[
\begin{align*}
\theta_o &= 0 \\
1 + \exp \left( \frac{nF}{RT}(E - E_p) + 2G \frac{\phi(\theta)}{\theta} \theta_o - G\phi(\theta) \right) &= 0 \\
1 + \exp(X) &= 1 + \exp(-X) \\
\end{align*}
\]  
(8)

From the equations 1 and 8, the current can be expressed with the \( X \) variable:

\[
i(X) = \frac{n^2F^2\text{Av}\Gamma_{\text{max}} \theta}{RT} \frac{\exp(X)}{(1 + \exp(X))^2 - 2B\exp(X)} 
\]  
(9)

With \( B = G\phi(\theta) \) an interaction constant, and \( |B| < 2 \).

To finalize the regression function, \( X \) needs to be expressed as a function of the applied potential. By combining expressions 7 and 8:

\[
X = \frac{nF}{RT}(E - E_p) + 2G \frac{\phi(\theta)}{\theta} \theta_o - G\phi(\theta) = \frac{nF}{RT}(E - E_p) + 2B \frac{\theta_o}{\theta} - B
\]

\[
X = \frac{nF}{RT}(E - E_p) + \frac{2B}{1 + \exp(-X)} - B
\]  
(10)

Equation 10 shows that the variable \( X \) is recursive, i.e. it is defined as a function of itself.

### 3.2. Approximation of the \( X \) variable and definition of GLI function

To circumvent the problem of recursion, an approximation of the expression \( 2B \frac{\theta_o}{\theta} \) included in the equation 10 must be determined. To achieve this, a set of simulations with different values of \( B \) was performed. To illustrate simulations, Figure 1A represents the dimensionless simulated current
\[ \Psi = \left( \frac{n^2 F^2 \ Av \ \Gamma}{RT} \right)^i \]
of a SAM using \( B = 0.8 \), and Figure 1B the different components of the variable \( X \)
\[ \begin{align*}
X &= X_a + X_b + X_c; \\
X_a &= \frac{2B}{RT} \left( E - E_p \right) \\
X_b &= \frac{2B}{1 + \exp(-X)} \\
X_c &= B
\end{align*} \]

By definition, component \( X_b \) is a perfect sigmoid as a function of \( X \) but not as a function of \( E \). However, by a
moderate approximation of \( X \) in \( X_b \) expression, it is possible to fit \( X_b \) according to \( E \) by a generalized
sigmoid (Figure 1C):
\[ X_{b_{in}} = \frac{2B}{1 + \exp\left(-\frac{nF}{RT} \left( E - E_p \right) \lambda \right)} \quad \text{with} \quad X = \frac{nF}{RT} \left( E - E_p \right) \lambda \]  

\( \lambda \) parameter allows modulation of the slope at the inflection point and is only \( B \) dependant (Figure 1D), the
inverse of \( \lambda \) being perfectly proportional to \( B \):
\[ \frac{1}{\lambda} = 1 - \gamma \cdot B \quad \text{with} \quad \gamma = 0.399 \pm 0.001 \approx 0.4 \quad \text{when} \quad B < \left| 2 \right| \]  

To conclude, the GLI function can be expressed as a function of the applied potential by combining the
equations 9, 10, 11 and 12:
\[ i(E) = \frac{n^2 F^2 \ Av \ \Gamma \exp(X)}{RT \left( 1 + \exp(X) \right)^2 - 2B \exp(X)} \]
\[ \begin{align*}
X &= \frac{nF}{RT} \left( E - E_p \right) + \frac{2B}{1 + \exp\left(-\frac{nF}{RT} \left( E - E_p \right) \right)} - B \\
\end{align*} \]

\[
\begin{align*}
\gamma &= 0.4 \\
B &= G \phi(\theta) \quad \text{and} \quad |B| < 2
\end{align*} \]

And \( E_p, B \) and \( \Gamma \) (\( \Gamma = \Gamma_{\max} \)) are the unknown parameters of the fit

3.3. Determination of characteristic parameters of fitted CVS

Peak potential (\( E_p \)) is directly determined by the fit of experimental CVs:
\[ E_p \pm \sigma_{E_p} \]  

Peak intensity (\( i_p \)) can be extracted easily with the equation 13:
\[ i_p = i(E = E_p \text{ or } X = 0) = \frac{n^2F^2Av}{RT} \frac{\Gamma}{4 - 2B} \]  
\[ \sigma_p = \frac{n^2F^2Av}{RT} \sqrt{\left(\frac{1}{4 - 2B} \sigma_r\right)^2 + \left(\frac{2\Gamma}{(4 - 2B)^2} \sigma_b\right)^2} \]  

An approximation of FWHM can be deduced at \( \frac{i_p}{2} \) from equations 9 and 11 (see Appendix):

\[
\text{FWHM} = E_+ - E_- = 2 \frac{RT}{nF\lambda} \ln \left(\frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}}\right) = 2 \frac{RT}{nF} \left(1 - \gamma B\right) \ln \left(\frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}}\right)
\]

\[
\sigma_{\text{FWHM}} = 2 \frac{RT}{nF} \gamma \ln \left(\frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}}\right) + \frac{2(1 - \gamma B)}{(B - 4)^2} \left(\frac{B - 2}{B - 4}\right)^2 \left(1 + \frac{B - 2}{B - 4}\right) \left(1 - \frac{B - 2}{B - 4}\right) \sigma_b
\]

It is noteworthy that the GLI function gives access to some physico-chemical characteristics, unlike usual functions. With GLI function, surface coverage \( \Gamma \) and the interaction constant \( B \) are directly determined by the fit of experimental CVs:

\[ \Gamma \pm \sigma_{\Gamma} \quad \text{and} \quad B \pm \sigma_B \]  

4. Discussions

4.1. Ideal case

In the ideal case of immobilized electroactive centers (i.e. Langmuir model where all adsorption sites are equivalent and there is no intersite interaction), CVs shape are independent of the surface coverage and can be described by the algebraic equation formulated by Laviron [5]:

\[
i(E) = \frac{n^2F^2Av \Gamma}{RT} \frac{e^{\left(\frac{nF(E - E_v)}{RT}\right)}}{\left(1 + e^{\left(\frac{nF(E - E_v)}{RT}\right)}\right)^2}
\]

It is worth noting that for \( B = 0 \) (i.e. ideal case), the GLI function (Eq. 13) simplifies into an expression similar to the Laviron case (Eq. 18), i.e. the ideal case is natively a particular case of the GLI function.
4.2. GLI function vs. generalized lateral interactions model

To test the reliability of the GLI function in controlled conditions, a set of CVs was simulated in the case of phase segregation \( \phi(\theta) > \theta \) for two significant values of G parameter (i.e. G = ± 1) and fitted with the GLI function in order to extract characteristic parameters \( E_p, I_p, \text{FWHM}, \Gamma \) and \( G\phi(\theta) \). These parameters were then plotted versus the theoretical surface coverage \( \Gamma \) and compared to the theoretical values from the GLI model (Figure 2 and Figure 3).

As expected, the agreement between the simulated voltammetry peaks and the fitted curves in the two borderline cases of phase segregation provides evidence of the usefulness of the GLI function. All characteristic parameters fit with the theoretical ones in a wide range of surface coverage. Even the approximate function (Eq. 16) to calculate the FWHM allows, compared to a graphic determination, obtaining a good estimate of this parameter. It is noteworthy a good agreement of the regression analysis in the difficult case of narrow peaks (i.e. G > 0) compared to previous works where sharp peaks and discontinuities are observed.

4.3. GLI function vs. experimental voltammograms

We confronted the GLI function to experimental CVs of TTF SAMs [15]. This well-known system presents two successive oxidations, and thus makes the linearization procedure more difficult.

The major difficulty which can disturb the performance of curve fitting processes is the non-predictable trend of experimental CV baseline. To circumvent this problem, a monotonic polynomial function (i.e. \( g(x) = \pm \sum_{i=0}^{n} |a_i x^i| \)) was added to functions.

An important hypothesis is that charges involved must be conserved during the two successive oxidations of TTF molecules, so this point is imposed on the fitting procedure.

Figure 4A shows that GLI function is able to extract all components of the experimental CV. As expected, the charge of each oxidation process is conserved and the two successive voltammetry peaks are consistent with GLI model and present repulsive and attractive interactions, respectively.

The regression analysis from GLI function was then compared to previous work performed with a Generalized Gaussian function (GG function) on CV of TTF SAM (Figure 4B) [14]. As anticipated, no difference is observed on the first oxidation step because the peak is rather broad. By contrast, the narrowness of the second peak allows assessment of the quality of the regression analysis via the GLI
function, compared to GG function which produces a sharp peak not consistent with raw data. This
difference is because GLI function is not empiric but is based on a model that leads the voltammetric shape.
To support that, we confronted the GLI function to experimental CVs of TEMPO SAMs, a redox-responsive
material well-known for its strong attractive interactions (G>0) between oxidized species in CH$_2$Cl$_2$ [6]. Figure
4C and 4D show a real difference between the two fitting curves despite the correlation coefficients being
rather close. Once again, the GLI function fits the shape of the experimental curves and allows extracting the
characteristic parameters of CVs, with good accuracy.
Note that with good initial values of (E$_p$, B and $\Gamma$), the fitting procedure from GLI function rapidly converges
and reaches an absolute minimum.

5. Conclusion
Derived from the GLI model, a regression function has been developed and tested on simulated and raw
data in order to fit voltammetric peak of self-assembled monolayers for extracting easily characteristic
parameters such as the full width at half maximum, the peak potential and the peak current.
Furthermore, the GLI function gives access to some physico-chemical characteristics such as surface
coverage and interaction constant, unlike the usual functions.
This work reinforces the idea that a regression function derived from a theoretical model is more suitable to
fit raw data than an empirical function like Gaussian, Generalised Gaussian, Lorentzian and Extreme
functions.

Acknowledgments
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Appendix

FWHM approximation

FWHM cannot be extracted at $\frac{i_p}{2}$ from equation 13. An alternative is to solve equation 9 according to variable $X$, and use equation 11 to obtain an approximation of FWHM.

$$\Psi_{FWHM} = \frac{\Psi_p}{2} = \frac{1}{8 - 4B} \left(\frac{\exp(X)}{(1 + \exp(X))^2 - 2B \exp(X)} \right) = \frac{\chi}{(1 + \chi)^2 - 2B\chi}$$

with $\chi = \exp(X)$

$$\chi^2 + 2(B - 3)\chi + 1 = 0$$

$$\begin{align*}
\chi_+ &= (3 - B) + \sqrt{B^2 - 6B + 8} \\
\chi_- &= (3 - B) - \sqrt{B^2 - 6B + 8}
\end{align*}$$

and

$$\begin{align*}
X_+ &= \ln\left((3 - B) + \sqrt{B^2 - 6B + 8}\right) \\
X_- &= \ln\left((3 - B) - \sqrt{B^2 - 6B + 8}\right)
\end{align*}$$

The symmetry of $I$ - $E$ curve at $E = E_p$ (i.e. $X = 0$) imposes that $X_+ = -X_-$ and leads to:

$$X_+ = -X_- = \ln\left((3 - B) + \sqrt{B^2 - 6B + 8}\right) = \ln\left(\frac{1 + \frac{B - 2}{\sqrt{B - 4}}}{1 - \frac{B - 2}{\sqrt{B - 4}}}\right)$$

(E) and $E_-$(can then be expressed from the approximation of equation 11 and the equations 12 and A1:

$$\begin{align*}
X_+ &= \ln\left(\frac{1 + \frac{B - 2}{\sqrt{B - 4}}}{1 - \frac{B - 2}{\sqrt{B - 4}}}\right) = \frac{nF}{RT}(E_+ - E_p)\lambda \\
X_- &= -\ln\left(\frac{1 + \frac{B - 2}{\sqrt{B - 4}}}{1 - \frac{B - 2}{\sqrt{B - 4}}}\right) = \frac{nF}{RT}(E_- - E_p)\lambda
\end{align*}$$

$$\Rightarrow$$

$$\begin{align*}
E_+ &= \frac{RT}{nF\lambda} \ln\left(\frac{1 + \frac{B - 2}{\sqrt{B - 4}}}{1 - \frac{B - 2}{\sqrt{B - 4}}}\right) + E_p \\
E_- &= -\frac{RT}{nF\lambda} \ln\left(\frac{1 + \frac{B - 2}{\sqrt{B - 4}}}{1 - \frac{B - 2}{\sqrt{B - 4}}}\right) + E_p
\end{align*}$$

(A2)

From equation A2, FWHM and its standard deviation can be deduced:
FWHM = \( E_+ - E_- = 2 \frac{RT}{nF\lambda} \ln \left( \frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}} \right) = 2 \frac{RT}{nF} (1 - \gamma B) \ln \left( \frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}} \right) \)  

\[ \sigma_{\text{FWHM}} = 2 \frac{RT}{nF} \left[ \gamma \ln \left( \frac{1 + \sqrt{\frac{B - 2}{B - 4}}}{1 - \sqrt{\frac{B - 2}{B - 4}}} \right) + \frac{2(1 - \gamma B)}{(B - 4)^2} \sqrt{\frac{B - 2}{B - 4}} \left( 1 + \sqrt{\frac{B - 2}{B - 4}} \right) \right]^{\frac{1}{2}} \]  

(A3)
References

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FIGURES
Figure 1.
(A) Dimensionless simulated current of a SAM vs. \((E-E_0')\) using \(B = G\phi = 0.8\) (and \(S\phi = -0.8\)).
(B) Decomposition of the different components of the variable \(X\) (\(X = X_a + X_b + X_c\)) vs. \((E-E_0')\) in the case of the simulation performed in A).
(C) \(X_b\) component (red) and fitted curve of \(X_b\) component (blue) using a generalized sigmoid vs. \((E-E_0')\) in the case of the simulation performed in A).
(D) \(\lambda\) parameter of generalized sigmoids extracted from a set of simulations vs. \(B\) parameter.
Figure 2.
Evaluation of GLI function peak fitting on simulated CVs from GLI model in the case of full reversible reaction
\( \frac{k}{v} \rightarrow \infty \), \( n = 1 \), \( k = 1000 \text{ s}^{-1} \), \( E_0 = 0.500 \text{ V} \), \( T = 293 \text{ K} \), \( v = 0.1 \text{ V.s}^{-1} \), \( A = 0.2 \text{ cm}^2 \) and \( \Gamma_{\text{max}} = 5 \times 10^{-10} \text{ mol.cm}^{-2} \)
in the case of \( G = +1 \) and \( S = -1 \).
(A) (lines) simulated CVs with different values of surface coverage (\( \theta \)) and phase segregation (\( \phi \)), and
(dotted lines) fitted curves using GLI function.
Comparison of (---) theoretical, (●) fitted and (●) graphically estimated parameters: (B) surface coverage, (C) B parameter, (D) peak potential, (E) full width at half maximum and (F) peak intensity
Figure 3.
Evaluation of GLI function peak fitting on simulated CVs from GLI model in the case of full reversible reaction \( \frac{k}{v} \to \infty \), \( n = 1 \), \( k = 1000 \, \text{s}^{-1} \), \( E'_0 = 0.500 \, \text{V} \), \( T = 293 \, \text{K} \), \( v = 0.1 \, \text{V.s}^{-1} \), \( A = 0.2 \, \text{cm}^2 \) and \( \Gamma_{\text{max}} = 5 \times 10^{-10} \, \text{mol.cm}^{-2} \) in the case of \( G = -1 \) and \( S = +1 \).

(A) (lines) simulated CVs with different values of surface coverage (\( \theta \)) and phase segregation (\( \phi \)), and (dotted lines) fitted curves using GLI function.

Comparison of (---) theoretical, (●) fitted and (●) graphically estimated parameters: (B) surface coverage, (C) B parameter, (D) peak potential, (E) full width at half maximum and (F) peak intensity.
Figure 4.
Evaluation of GLI function peak fitting on experimental CV of TTF [15] and TEMPO [6] SAMs in 0.1 M Bu$_4$NPF$_6$/CH$_2$Cl$_2$, at 0.1 V.s$^{-1}$ and 293 K, and comparison to Generalized Gaussian (GG) function peak fitting.

(A) Electrochemical parameters deduced of two peaks fitting using GLI function on TTF SAM:
\[
\Gamma = 0.89 \times 10^{-10} \text{ mol.cm}^{-2}, B_1 = -0.659, E_{p1} = 0.231 \text{ V}, \text{FWHM}_1 = 0.126 \text{ V}, i_{p1} = 1.28 \times 10^{-6} \text{ A, } E_{p2} = 0.547 \text{ V, } B_2 = 0.516, \text{FWHM}_2 = 0.062 \text{ V, } i_{p2} = 2.29 \times 10^{-6} \text{ A}
\]

(B) Electrochemical parameters deduced of two peaks fitting using GG function on TTF SAM:
\[
\Gamma = 0.90 \times 10^{-10} \text{ mol.cm}^{-2}, E_{p1} = 0.231 \text{ V, FWHM}_1 = 0.122 \text{ V, } i_{p1} = 1.31 \times 10^{-6} \text{ A, E}_{p2} = 0.547 \text{ V, FWHM}_2 = 0.057 \text{ V, } i_{p2} = 2.29 \times 10^{-6} \text{ A}
\]

(C) Electrochemical parameters deduced of two peaks fitting using GLI function on TEMPO SAM:
\[
\Gamma = 4.68 \times 10^{-10} \text{ mol.cm}^{-2}, B_1 = 1.071, E_{p1} = 0.4940 \text{ V, FWHM}_1 = 0.037 \text{ V, } i_{p1} = 19.26 \times 10^{-6} \text{ A}
\]

(D) Electrochemical parameters deduced of two peaks fitting using GG function on TEMPO SAM:
\[
\Gamma = 4.77 \times 10^{-10} \text{ mol.cm}^{-2}, E_{p1} = 0.494 \text{ V, FWHM}_1 = 0.029 \text{ V, } i_{p1} = 21.73 \times 10^{-6} \text{ A}
\]
Graphical abstract
**Highlights**

- A function derived from a theoretical model fits adsorption voltammetric peak.
- This approach is dedicated to SAMs where interactions between electroactive centers are possible.
- The fitting procedure is easy to implement and gives access to characteristic parameters of a voltammetric peak (i.e. $E_p$, $i_p$, FWHM, $\Gamma$, and interaction parameter).