A Microelectrode-Cell Sensor Model for Real Time Monitoring

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Abstract-- In this paper the application of a cell-microelectrode model to cell biometry experiments is proposed, using the cell-electrode area overlap as main parameter. The model can be applied to cell size identification, cell count, and their extension to cell growth and dosimetry protocols. Experimental results using AA8 cell line are presented, obtaining promising results.

Keywords- Microelectrode; ECIS; bio-impedance; impedance sensor; cell culture; dosimetry.

I. Introduction

Many biological parameters and processes can be sensed and monitored using its impedance as marker [1-5], with the advantage of being a non-invasive and relatively cheap technique. Cell growth and activity, changes in cell composition and shape, or in cell location are examples of how processes can be detected with microelectrode-cell impedance sensors [6-9]. Among Impedance Spectroscopy (IS) techniques, Electrical Cell-substrate Impedance Spectroscopy (ECIS) [7,8], based on two-electrode setups, allows the measure of cell-culture impedances and the definition of the biological nature (material, internal activity, motility and size) of a kind of cell and its relationship with the environment [11]. One of the drawbacks of ECIS technique is the need of efficient models to decode the full system electrical performance composed by the electrodes, medium and cells. Several works have been developped in this field. In [8], magnitude and phase impedance are deduced from electric field equation solution at the cell-electrode interface, giving a three parameter based model. h, the cell-electrode distance, R_b, cell-to-cell barrier resistance and r_{cell}, cell radius. In [9,10], finite element simulation (FEM) are executed to solve electrical field considering the whole structure. This method gives one parameter model (R_{gap}) to describe the gap or cell-electrode region resistance. In both, the model considers cells are in confluent phase [7] or a fixed area over the electrode [9]. The latest was extended in [10] to several cell sizes, allowing to define the cell-electrode covered area as the main model parameter. In this work is considered a model extension of $R_{\mbox{\scriptsize gap}}$ based model, to incorporate the variable cell-microelectrode area overlap [10]. Impedance sensor sensitivity curves based on the cell size and density will be presented and applied to measure the growth-tax in cell-cultures and to describe cell toxicity experiments.

In this paper, section II resumes the electrode solution model for cell-electrode characterization. The process to extract practical models is included at section III, illustrating the simulations on a simplified system leading to cell size Paula Daza Dpt. Cell Biology Biology Faculty, Seville University Av. Reina Mercedes sn, 41012, Seville, Spain pdaza@us.es

detection. Section IV relies on real time cell culture monitoring and its application to dosimetry experiments. Conclusions will be highlighted at section V.

II. ELECTRODE-ELECTROLYTE MODEL

The impedance of electrodes in ionic liquids has been rather extensively investigated. An excellent review can be found at [6]. The main componets describing the electrical performance of an electrode metal inside a solution are four: the double layer capacitance, C_I , the current flowing through the electrified interface will encounter a resistance R_{ct} caused by the electron transfer at the electrode surface and Warburg impedance Z_W due to limited mass diffusion from the electrode surface to the solution. The electron transfer resistance R_{ct} is in series with the mass diffusion limited impedance Z_W . As the current spreads out to the bulk solution, the electrode has a solution conductivity determined by series resistance, represented as spreading resistance R_S in the equivalent circuit. These four parameters depends on technology, medium and geometry.

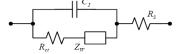


Figure 1. Equivalent circuit of electrode-solution interface. $C_{\rm I}$ is the double layer capacitance. Faradic impedance includes $Z_{\rm w}$, the Warburg impedance and $R_{\rm ct}$, the charge-transfer resistance. $R_{\rm s}$ is the spreading resistance.

III. CELL-ELECTRODE MODEL

The Fig. 2 illustrates a two-electrode impedance sensor useful for ECIS technique: e_1 is the sensing electrode and e_2 the reference one. Electrodes can be manufactured in CMOS process with metal layers [9] or using post-processing steps [13]. The cell location and size on e_1 top must be detected.

The model in Fig. 3 considers the sensing surface of e_1 could be total or partially filled by cells. For the two-electrode sensor in Fig. 2, e_1 is the sensing area A, $Z(\omega)$ is the impedance by unit area of the empty electrode (without cells on top). When e_1 is partially covered by cells in a surface A_c , $Z(\omega)/(A-A_c)$ is the electrode impedance associated to non-covered area by cells, and $Z(\omega)/A_c$ the impedance of the covered area. R_{gap} models the current flowing laterally in the electrode-cell interface, which depends on the electrode-cell distance at the interface (in the range of 15-150nm). R_s is the spreading resistance through the conductive solution. For an empty electrode, the impedance model $Z(\omega)$ is represented by the circuit in Fig. 1. It has been considered for e_2 the model in Fig 3a, not covered by cells. The e_2 electrode is commonly large

and ground connected, being its resistance small enough to be rejected. Figure 4 represents the impedance magnitude, Z_c , for the sensor system in Fig. 2, considering that e_1 could be either empty, partially or totally covered by cells. The parameter ff, called $fill\ factor$, can be zero for $A_c=0$ ($e_1\ electrode\ empty$), and 1 for $A_c=A$ ($e_1\ electrode\ full$). It is defined $Z_c\ (ff=0)=Z_{nc}$ as the impedance magnitude of the sensor without cells.

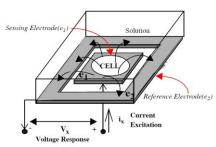


Figure 2. Two electrodes for ECIS: e_1 (sensing) and e_2 (reference). AC current i_x is injected between e_1 - e_2 , and voltage response V_x is measured.

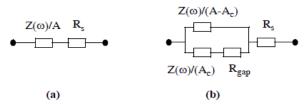


Figure 3. Proposed model for the an electrode-solution-cell model with area A, uncovered with cells (a) and covered and area A_c (b).

The relative changes at impedance magnitude, defined as,

$$r = \frac{Z_c - Z_{nc}}{Z_{nc}} \tag{1}$$

inform more accurate from these variations, being r the change of impedance magnitude for the two-electrode with cells (Z_c) with respect to the system without them (Z_{nc}). The graphics of r versus frequency is plotted in Fig. 5, for a cell-to-electrode coverage ff from 0.1 to 0.9 in steps of 0.1, using a R_{gap} =90 k Ω . The size of the electrode is 32x32 μm^2 [9,10]. It can be identified again the frequency range where the sensitivity to cells is high at 100kHz, represented by r increments. For a given frequency, each normalized impedance value of r can be linked with its ff, being possible the cell detection and estimation of the covered area A_c . Even more, area covered can be intrepreted as consequence than two o more cells, allowing cell count for a given cell size.

From Fig. 5, it can be deduced that models of electrode-cell electrical performance can be used to derive the overlapping area in cell-electrode systems, useful for biological studies. It can be observed how the curve fits well with the frequency range, placing the maximum r value around 100 kHz, as predicts the FEM simulations [9, 10]. A value of $R_{\rm gap}$ = 90k Ω was selected for this curve, representing a maximum value of the r curve with ff=0.69, which represents the ratio (A_c/A), for a cell size of 30µm diameter represented at figure obtained using FEM simulations [10]. Impedance sensor curves at figures 4 and 5 were obtained using SpectreHDL [15] mixed-

mode simulator, with Analog Hardware Description Language (AHDL) for circuits in Fig. 3. An advantage of using AHDL models is the possibility of including non-linear performance of circuit elements, in our case, the frequency squared-root function at the Warburg impedance.

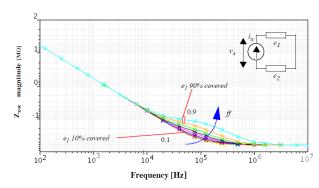


Figure 4. Impedance evolution when fill factor increases $32 \times 32 \mu m^2$.

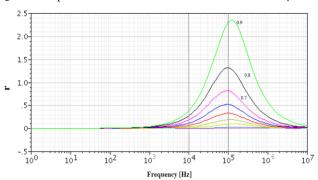


Figure 5. Normalized impedance r versus frequency derived from Fig. 4. Curves correspond to ff in the range of 0.1 (near empty) to 0.9 (near full).

IV. CELL CULTURE APPLICATIONS

A: Electrode Model

The proposed model based in Fig 3 has three main parameters: the electrode area (A), the fill-factor (ff), and the resistance of the gap region (Rgap). Technology data were included and simulation results obtained to model a commercial electrode: 8W10E, from Applied Biophysics [12]. It is composed by eight wells; each one contains ten circular gold microelectrodes, with 250µm diameter. Ten sensing electrodes, in paralell, were used for e₁ and only one common reference electrode, much larger than sensing ones. Figure 6 represents the normalized impedance r expected for these electrodes, for R_{gap} =22k Ω , if fill factor changes from electrodes without cell on top (ff=0.1) to near those fully covered (ff=0.9). Values of R_{gap} can be used to match the models to observed performance. In Fig. 7, R_{gap} values were changed for ff=0.9, observing large r changes. Finally, it was also modifyed the electrode area for $R_{\text{gap}}\!\!=\!\!22k\Omega$ and ff=0.9, showing the results at Fig. 8. It can be observed that optimal working frequency is near the proposed by the electrode factory (around 4 kHz), and that electrode area covered by cells can be approximated by using the fill factor parameters. The performance curves obtained before can be used to fit experimental results to proposed model and find relevant biometric characteristics.

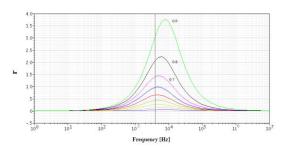


Figure 6. Obtained curves for r vs frequency, for ff \in [0.1,0.9] and R_{gap} = $22k\Omega$, using 8W10E electrodes.

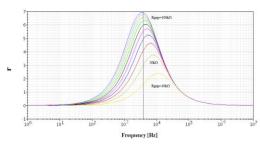


Figure 7. Obtained curves for r vs frequency, for $R_{gap} \in [10k\Omega, 100k\Omega]$ in steps of $10k\Omega$, for ff=0.9, using 8W10E electrodes.

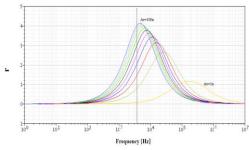


Figure 8. Obtained curves for the r vs frequency, for R_{gap} =22k Ω and ff=0.9, for differents electrode areas (1n to 100n). 49n (49.10⁻⁹m²) corresponds to a circular electrode with a 250 μ m diameter.

B: Cell growth

In Fig. 9a it is shown the growth curve obtained by us during seven days using 8W10E sensors with a similar setup in [8]. AA8 cells for chinese hamster were seeded initially, in an approximated number of 5000. The impedance range is around 1220Ω (380Ω-1600Ω). Considering an initial cell number of 5000 very low, we take the initial impedance as due to no-cell impedance value (Znc). At t=6000 min, the medium was changed, and the confluent phase was achieved at t = 8500 min approximately. The maximum experimental value given from eq. (1) is around r = 3.1, as illustrates Fig. 9b. We consider in our model that the electrodes are aproximatelly fully covered by cells for ff=0.9, the value of R_{gap} that better fits is $22k\Omega.$ System response corresponds to r-values illustrated in Fig. 6. From these curves, it can be obtained the fill factor at different times. Table I summarized the relative normalized impedance values r at several times. Using Fig. 6 for the sensor response, fill factor is calculated at every instant. For a well area of 0.8 cm², the maximum cell number goes from 0.8×10^6 to 1.6×10^6 Number of cells, n_{cell}, in Table I, is obtained from 0.8x10⁶ expected final cell number. A value of $Z_{\text{nc}}\!\!=\!\!380\Omega$ for r calculus in eq. (1) was considered.

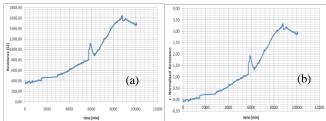


Figure 9. (a) Impedance evolution of the cell growth experiment. (b) The normalised impedance r evolution obtained.

Table Me: Cell Number (n_{cell}) obtained from impedance Z_c measure in Fig. 9, and using the r curves proposed for 8W10E sensors.

. 9, AND USING THE F CURVES PROPOSED FOR 8 W TUE SENSORS.								
t (min)	r	ff	n_{cell}					
0	0	-	5000					
500	0.024	0.020	18000					
1000	0.050	0.050	44000					
1500	0.072	0.070	63000					
2000	n.a.	n.a.	n.a.					
2500	n.a.	n.a.	n.a.					
3000	0.374	0.362	322000					
3500	0.437	0.395	351000					
4000	0.615	0.475	422000					
4500	0.777	0.530	471000					
5000	0.903	0.581	516000					
5500	1.033	0.602	535000					
6000	1.074	0.620	551000					
6500	1.507	0.710	631000					
7000	1,970	0.775	689000					
7500	2,353	0.810	720000					
8000	2,837	0.860	764000					
8500	3.113	0.890	791000					
9000	3.134	0.900	800000					
9500	3.010	0.875	778000					
10000	2,857	0.864	768000					

C: Dosimetry

Experiments to characterize the influence of some drugs in cell growth were done. The objective is to proof that proposed model allows counting cell number at different dosis. It was considered the AA8 cell line and as drug, six different doses of MG132 for growth inhibition (from $0.2\mu M$ to $50\mu M$). After 72 hours normal cell growth, the medium was changed and the drug added at different doses: 0.2, 0.5, 1, 5, 10 and $50\mu M$ for wells 3 to 8 respectively. Well 2 is the control.

Measured impedances for the 8 wells are at Fig. 10, for 4kHz working frequency. At the end of the experiment can be observed that impedance decreases as drug dosis increases. Control (W2) is full of cells with the maximum impedance, while maximum dosis (W8) has the lowest resistance, at the botton. The black line (W1) represents the electrode-solution impedance. After the medium change (t=4000min), it is observed a decreasing impedance below the initial baseline level (400 Ω) that we cannot explain. Final impedance values at 8000min, Z_c , were considered, at Table II. From Z_{nc} and Z_c , rvalues are calculated in third columm. Using curves for r versus frequency in Fig. 6, ff estimated values from proposed model are obtained. The cell number at the end of the experiment was also count and shown at the last column for each well. Considering $ff_{max}=0.9$ for a measured cell number of 8.06×10^3 , the expected values for ff are calculated.

The same data are summarized at Table III for 2, 4 and 10 kHz frequencies respectively. The better agreement it is obtained at 4 kHz in fill factor (ff). It is observed that the impedance baseline, Z_{nc} , for r calculus decreases with frequency due probably to electrode impedance dependence.

For medium resistance (W1) and high drug concentrations wells (W6-W8), the resistance measured is below to Z_{nc} , so eq. (1) can no be applied for r calculation.

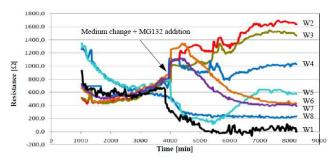


Figure 10. Impedance measure in dosimetry at 8 wells for 4kHz frequency. W1: Medium. W2: Control. W3: $0.2\mu M$. W4: $0.5~\mu M$. W5: $1~\mu M$. W6: $5~\mu M$. W7: $10~\mu M$ and W8: $50~\mu M$.

Table II: Experimental values for relative impedance (r) and fill-factor (ff) for Z_{nc} = 400 Ω . Frequency = 4 kHz.

ncell t=8000min Z_c , Z_n estimated expected 259.2 Medium 2 1631.7 3.1 0.90 0.900 8.06x105 3 1454.7 2.6 0.85 0.690 6.13x105 1030.6 4 1.5 0.72 0.610 5.41x105 625.8 0.5 0.44 0.410 3.60x10⁵ 5 6 417.4 0.05 0.037 0.036 3.20x104 406.8 0.015 0.016 0.024 2.10x104 99.6 < 0 0.005 4.00x103

Table III: Experimental values for relative impedance (r) and fill factor (ff) at various frequencies. Z_{NC} = 480 Ω , 400 Ω , and 315 Ω for 2, 4 and 10 kHz working frequency respectively.

	r (from Z_c and Z_{nc})			ff (from model)			
Well	2kHz	4kHz	10kHz	2kHz	4kHz	10kHz	ff expect.
1	-	-	-	-	-	-	Medium
2	2.43	3.1	3.76	0.98	0.90	0.90	0.900
3	2.18	2.6	3.24	0.94	0.85	0.88	0.690
4	1.17	1.5	2.21	0.82	0.72	0.82	0.610
5	0.21	0.5	0.84	0.32	0.44	0.44	0.410
6	-	0.05	-	-	0.037	-	0.036
7	-	0.015	-	-	0.016	-	0.024
8	-	-	-	-	-	-	0.005

V. DISCUSSION AND CONCLUSIONS

This work describes an area dependent model for cell-electrode systems and its application to measure and identify cells during cell culture protocols. A practical circuit for electrode-solution-cell simulation was employed, using an AHDL description for commercial electrodes, obtaining a good matching. Optimal measurement frequency was identified near 4 kHz. It was proposed the cell growth evolution study based on 8W10E electrode models. Curves obtained experimentally, allows the real time growth monitoring by fitting the $R_{\rm gap}$ parameter. An estimation of the number of cells was obtained by using sensor curves calculated from electrical model proposed. Dosimetry experiments reproduce similar conditions than cell growth, but in this case, it is added a growth inhibitor at different dosis. There is observed a decreasing impedance, below the baseline expected ($Z_{\rm nc}$) that we can not explain.

However, for the control and small drug dosis, impedance curves are perfectly aligned. It was fitted a proposed model with $R_{\rm gap}\!\!=\!\!22k\Omega$ to explain experiemental data. Deviations from data are over 10-20% in fill factor, more accurate for 4 kHz.

The deviations in fill factors measured are not small, being required to analize the influence of error sources to increase the system performance. First, Signal-to-Noise Ratio (SNR) should be increased at the setup. Second, proposed model has the advantage that need only one parameter (R_{gap}), versus other reported model using three parameters [8]. One parameter model makes easy to fit experimental data, but can introduce inaccuracy. The possibility to add more parameters to the model should be considered in the future.

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