1 Toward noninvasive monitoring of plant leaf water content

2 by electrical impedance spectroscopy

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21 resistance modelling the extracellular fluid can track the change in water content of a plant leaf while

22 it is drying at room temperature. On the other hand, impedance data might provide information

23 about the leaf tissue structure at small inter-electrode distances.

24 Keywords: Plant electrical properties; electrical impedance spectroscopy; plant water content.

25 **1. Introduction**

26

Water resources are becoming scarce in most countries around the world. This makes the task of allocating those resources for agricultural production increasingly difficult (Kalyanaraman et al., 2022). However, plant hydration is one of the main factors that determines its wellbeing and yield hence efforts must be made to establish appropriate watering schedules. Sensing technologies and algorithms are currently being developed to help in detecting plant water stress (Elsherbiny et al., 2022), which combined with the experience of farmers constitute a powerful tool for establishing irrigation needs and optimizing water management.

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35 Hydric stress occurs mainly during the drought period and manifests as visible color and structural 36 changes -decrease in turgor, for example. One way to assess the hydric status of a plant is to measure 37 its relative water content (RWC) (Zimmermann et al., 2008). The gold standard method (Smart and 38 Bingham, 1974) relies on mass measurements at different leaf sample hydration states: fresh, water-39 saturated (full turgor) and oven-dried. The process is thus time-consuming -4 to 24 h of water 40 saturation and 24 h of desiccation in the oven, typically– even if one obviates the drying process. It 41 also requires cutting the leaves specifically, adding up to the complexity of the process. Moreover, 42 this method does not allow measuring leaf RWC in vivo or in situ. Therefore, noninvasive methods 43 have been proposed that involve contact and non-contact solutions. Near-infrared spectroscopy 44 (NIRS) is an indirect measurement technique that correlates to RWC to an acceptable degree (Hunt 45 et al., 1987; Torres et al., 2019) and can provide fast in situ estimates. However, the cost of most NIRS 46 equipment is currently very expensive and limits its deployment as a daily use tool in crop 47 maintenance. Alternatively, turgor can be directly measured on the plant leaves (Zimmermann et al., 48 2008) with very high precision, and relatively low cost.

50 Electrical impedance measurement is an inexpensive, easy to build, noninvasive technique that 51 provides information about the composition of biological media like plants (Jócsák et al., 2019). A 52 sinusoidal excitation source (a current or a voltage) at a specific frequency is applied onto the 53 biological medium by means of two electrodes attached to its surface. This creates a current flow 54 inside the biological medium and a voltage response is developed and read either at the same 55 injecting electrodes (two-electrode technique) or at a second set of electrodes placed conveniently 56 (four-electrode technique). The magnitude of the current injected into the medium is usually very 57 well below 1 mA. Bathroom scales that provide information about a person's body composition use this technique, calculating the total body water content from a single frequency measurement and 58 59 empirical formulae to estimate other parameters like fat and muscle content.

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61 Electrical impedance spectroscopy (EIS) uses the same setup to perform a frequency sweep, yielding 62 an impedance spectrum that fits into an equivalent electrical circuit consisting of passive electrical 63 elements (i.e. resistors capacitors). These circuit elements provide information about the electrical 64 properties of the sampled biological medium: the cell membrane electrical capacitance and the 65 intracellular and extracellular water conductivities (Pethig and Kell, 1987). In this way, the plant's 66 physiological activity or structural changes can be tracked as long as it affects its electrical properties. 67 EIS has been used to investigate the water content of roots and stems in potatoes and alfalfa (Hayden 68 et al., 1969). More recently, it has been used to investigate the leaf water content (LWC) with two 69 small electrodes (Barbosa et al., 2022) and found a high correlation between LWC and the impedance 70 magnitude, but the use of two electrodes might have masked the identification of the circuit elements 71 as it was not possible to extract the cell capacitance data nor the intracellular resistance. In such two-72 electrode measurements, the electrode impedance constitutes an interfering quantity, a problem 73 remediated using the four-electrode technique. Other EIS applications include the plant's response

to environmental changes (Garlando et al., 2022), to different growth and stress conditions (ben
Hamed et al., 2016), and to different plant water status (Jamaludin et al., 2015). It has also been used
to track the response to nitrogen nutrition stress in tomatoes (Li et al., 2017), and the thermallyinduced structural changes in spinach (Watanabe et al., 2017), to name some examples.

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79 Data interpretation and comparison between studies is, to some extent, difficult, as the measurement 80 protocols are mostly ad hoc. The electrodes used were different in their construction -shape and 81 material- and in their degree of invasiveness -surface electrodes vs. inserted needle electrodes. 82 Additionally, the electrodes were placed at very dissimilar locations (leaves or stems, or a 83 combination), making the data interpretation difficult. For example, (Zhang and Willison, 1993) 84 inserted two thin needle electrodes into plant leaf samples -causing damage to the leaves- while 85 (Jamaludin et al., 2015) use two surface electrocardiogram electrodes attached to plant leaves to 86 measure the impedance in vivo. In both cases, it was expected that the electrode impedance interfere 87 with the leaf impedance at the lower frequency range (hundreds of hertz range) but in the work by 88 Zhang and Wilson the insertion reduced this problem. However, the observed values were in the 89 kiloohm to tens of kiloohm range. These values are orders of magnitude larger than those obtained 90 by (Comparini et al., 2020) at dc, where the electrodes were inserted into the stem of an olive tree 91 plant in vivo, yielding values in the ohms to hundred ohms range. Although in these studies the 92 measured values were quite dissimilar, they all indicated an increase in impedance with decreased 93 plant water content.

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Wide acceptance of EIS as a technique to estimate the RWC is limited by two factors: i) the high cost and bulk of most commercial impedance analyzers, and ii) the accuracy of the measurement. For benchtop studies, the necessary frequency range will usually go from around 10 Hz to 1 MHz, although some studies report promising results with lower frequency ranges. This is the case of ownbuilt portable impedance analyzers based on commercially available integrated circuits like the
AD5933 (Analog Devices, Inc.) used to detect fruit ripening (Ibba et al., 2021). (Jamaludin et al., 2015)
also used this device, but the results were inconclusive as the impedance spectrum exhibited low
frequency values exceeding the AD5933 specifications, preventing the fit to an equivalent circuit.
Nonetheless, from 70 kHz to 100 kHz, the impedance magnitude showed a significant correlation
with RWC.

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106 In this work, a EIS measuring system has been implemented that is based on a low-cost USB 107 acquisition and generation system -the Analog Discovery 2 (AD2)- and a simple analog front-end 108 (AFE). The proposed system performs four-electrode impedance measurements in the frequency 109 range from 100 Hz to 1 MHz. This setup avoids the influence of electrode polarization that usually 110 arises when using two electrodes to measure the impedance. Leaf samples of spinacea oleracea L. 111 were prepared and their impedance measured and fitted to an equivalent circuit. Tests were 112 performed to assess the influence of electrode positioning and the capability of the system to track 113 water loss during the spontaneous drying of the leaf samples.

114

115 **2.** Materials and methods

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In this section, a novel measurement system able to acquire the impedance spectrum of a leaf in the
100 Hz - 1 MHz range is described. It uses four surface electrodes to prevent the interference of the
electrode impedance.

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121 2.1 EIS measurement system

Figure 1A shows a block diagram of the proposed EIS system. It comprises an AD2 and a fullydifferential AFE. Briefly, the AD2 has two differential input channels, CH1 and CH2 (14-bit resolution, 100 MSPS sample rate and 30 MHz analog bandwidth), two 14-bit single-ended waveform generators V_1 and V_2 that can be arranged to create a fully-differential waveform generator -12 MHz maximum frequency– and a programmable power supply, fixed at ±5 V. This fully-differential architecture was chosen to avoid issues with interfering signals that usually manifest as unwanted common-mode signals superimposed onto the impedance voltage response.

130

131 A reference resistor $R_{\rm sh}$ and the plant leaf impedance $Z_{\rm x}$ form a voltage divider. V₁ and V₂ are two 132 synced sine waves of constant amplitude (1 V) that have a relative phase of 180°. V₁- V₂ is thus used 133 as the fully-differential excitation source. Electrodes I₁ and I₂ are the current injection electrodes and 134 E₁ and E₂ are the voltage pick-up electrodes –reading the voltage response from the plant. A₁ and A₂ 135 are two identical AD8066 op amps (Analog Devices, Inc.) configured as a non-inverting fully-136 differential amplifier with a differential gain $G_d = 3$ (Figure 1B) fixed by of 499- Ω resistors –as 137 specified by the op-amp manufacturer. At this gain, the AD8066 has a -3 dB bandwidth well exceeding 138 10 MHz hence its gain and phase deviations are expected to be negligible. However, the AD8066 has 139 a very high input impedance that avoids voltage-loading effects due to the high leaf electrical 140 impedance. In a voltage divider, the reference resistor is usually chosen to be of similar magnitude to 141 the unknown impedance thus $R_{\rm sh}$ is 30 k Ω , which falls within the expected leaf impedance range and 142 is used to measure the current flowing into the leaf. Assuming that the electrode impedance does not 143 surpass 10 k Ω , a range of Z_x from 10 k Ω to 100 k Ω will produce a current between 33 μ A and 13 μ A. 144 These are very low amplitude currents hence common-mode interfering signals may distort the 145 impedance spectrum. By using a fully differential architecture, these interfering signals should not 146 introduce significant deviations. Smaller currents can also be measured, i.e., when the leaf begins to 147 dry and its impedance surpasses 100 k Ω , but then the impedance spectra might be more susceptible 148 to electromagnetic interferences and/or noise. Channel 1 of the AD2 is used to read the voltage drop

(1)

(2)

149 at $R_{\rm sh}$ (V_{CH1}) and Chanel 2 measures that of $Z_{\rm x}$ (V_{CH2}).

150 By applying Ohm's law, the impedance magnitude $|Z_x|$ and phase $\angle Z_x$ will be

$$|Z_x| = R_{\rm sh} \frac{|V_{\rm CH2}|}{|V_{\rm CH1}|}$$

$$\angle Z_x = \angle V_{\rm CH2} - \angle V_{\rm CH1} \quad .$$

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154

Figure 1. A) Block diagram of the proposed EIS measurement system, showing its two main components: the implemented
AFE and the AD2, controlled by a laptop PC. B) Schematic of the fully-differential amplifier design used in A₁ and A₂, that is
based on two AD8066 op amps (OA1 and OA2).

158

An acquisition and control virtual instrument was developed in LabVIEW (National Instruments Corp.). V_{CH1} and V_{CH2} were digitally processed to compute the amplitudes and phases in (1) and (2) with LabVIEW's in-built tone detection block based on a fast-Fourier transform computation, making it unnecessary to demodulate the signals from V_{CH1} and V_{CH2} . In this work, a logarithmic frequency sweep of 101 points between 100 Hz to 1 MHz was used, but this can be easily reprogrammed as long as the frequency does not fall outside the 10 Hz to 10 MHz range (10 MHz is the maximum imposed by the AD2, but the AFE performance at such high frequency will be poor). The system was calibrated 166 with a 30 k Ω resistor whose reference magnitude and phase values were measured with a 4294A

167 impedance analyzer (Agilent Technologies) in the frequency range from 100 Hz to 1 MHz.

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169 2.2 Impedance spectrum data fitting

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171 Biological tissues exhibit an impedance spectrum with clearly distinguishable dielectric dispersion 172 regions, which are the frequency-domain counterpart of the time-domain relaxation processes. In 173 animals, the small cell sizes (typically up to 100 μ m) yield three dispersion regions: α , β , and γ . The β 174 dispersion occurs at a frequency specific to the tissue under test and is linked to the cell membrane. 175 The frequency range below and above this dispersion yields information about the intracellular and 176 extracellular ionic contents. Figure 2A shows the equivalent circuit of a single dispersion tissue that 177 models the possible current flow paths. There, Z_{CPE} is the membrane impedance modeled by a 178 constant phase element; R_{ECF} and R_{ICF} model the extracellular fluid and the intracellular fluid, 179 respectively. Figure 2B is an equivalent circuit derived from the Cole-Cole empirical equation:

180
$$Z(\omega) = R_{\omega} + \frac{R_0 - R_{\omega}}{1 + (R_0 - R_{\omega})(j\omega)^{\alpha} C} = R_{\omega} + \frac{\Delta R}{1 + (j\omega)^{\alpha} \Delta RC}$$
(3)

181 where $\Delta R = R_0 - R_{\infty}$, R_0 is the impedance at zero frequency and R_{∞} is the impedance at infinite 182 frequency. The models in Figure 2A and Figure 2B yield equivalent information about the single 183 dispersion. To derive (3) a constant-phase element (CPE) has been used and whose impedance is 184 described by

185
$$Z_{\rm CPE} = \frac{1}{(i\omega)^{\alpha} C},$$

186 where α and *C* are empirical constants that allow good data fitting. The CPE represents an imperfect 187 capacitor, or pseudo-capacitor. Here, *C* represents the cell-membrane pseudo-capacitance and α is 188 interpreted as a parameter accounting for the dispersion of cell sizes. α is an intrinsic tissue

(4)

- 189 parameter and is independent of the sample geometry –unlike resistors and capacitors– and has been
- 190 suggested as a parameter to classify tissues or tissue status.



Figure 2. A) Equivalent circuit of an animal cell with a simple dispersion, as derived from the possible current paths. B)
Equivalent circuit derived from the Cole-Cole equations. C) Cole-Cole impedance plot showing a depressed semicircle
corresponding to either of the equivalent circuits.

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Figure 3. Equivalent circuits for plant tissues: A) Single-shell model, where P stands for plasma (cell wall and membrane);
B) Hayden model, where CW stands for the cell wall, CP for the cell cytoplast. C) Double-Shell model, where TP stands for
the vacuole tonoplast and IV for the inner vacuole. D) Cole-Cole impedance plot showing a double semicircle, as expected
in a double-shell model.

201

In plants, the cell size is larger (typically up to 300 µm) and the dispersion regions move toward 202 203 lower frequencies compared to animal cells. Additionally, plants exhibit a somewhat different 204 structure as they have a cell wall besides a cell membrane. Besides, plant cells have large vacuoles 205 that can occupy more than 80% of the total cell volume. The result is that, when drawing the Cole-206 Cole plots, they may exhibit a double arc, and the fit to single-dispersion circuits is not good. Instead, 207 circuits that are more complex are used: the single-shell model, the Hayden model, or the double-208 shell model (Prasad and Roy, 2020). The latter exhibits a double arc in the Cole-Cole plot, while the 209 former exhibits only one, as in animals. Double arcs appear in plant protoplasts (Asami et al., 1996),

210 roots (Repo et al., 2005) and leaves (Zhang and Willison, 1993). However, in some cases, only one

211 visible arc is displayed and the single-shell or Hayden equivalents are a suitable choice. One downside

of such circuits is that, usually, the plotted arcs are displaced downwards, meaning that a CPE might

213 be a better choice than capacitors.

In this work, a modified double-shell model was used, where the capacitors in Figure 3C were replaced with CPEs as shown in Figure 4. These CPEs are Z_P , modelling the plasma impedance, and Z_{TP} , modelling the tonoplast impedance. Their expressions are

217
$$Z_{\rm P} = \frac{1}{C_{\rm P} \left(j\omega\right)^{\alpha_{\rm P}}} \tag{5}$$

218
$$Z_{\rm TP} = \frac{1}{C_{\rm TP} \left(j\omega\right)^{\alpha_{\rm TP}}}.$$

(6)

219

Figure 4. Modified double-shell model where the capacitors are replaced with CPEs: Z_P is the plasma CPE and Z_{TP} is the tonoplast CPE.

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As with the series Cole-Cole equations, each CPE associates with a set of resistances, modeling a specific relaxation process. At low frequencies, Z_{TP} is an open circuit and the associated plasmatic membrane characteristic frequency f_P will be

226
$$f_{\rm P} = \frac{1}{2\pi \left[\left(R_{\rm CP} + R_{\rm CW} \right) C_{\rm P} \right]^{\frac{1}{\alpha_{\rm P}}}}.$$
 (7)

227 On the other hand, at high frequencies, Z_P is shorted, and the tonoplast characteristic frequency f_{TP} 228 will be

229
$$f_{\rm TP} = \frac{1}{2\pi \left[\left(R_{\rm IV} + R_{\rm CP} \parallel R_{\rm CW} \right) C_{\rm TP} \right]^{\frac{1}{\alpha_{\rm TP}}}}.$$

In the case of a modified single-shell model where C_p is replaced with a CPE modelled by (5), the

(8)

232 characteristic frequency $f_{\rm c}$ will be

233
$$f_{\rm c} = \frac{1}{2\pi \left[\left(R_{\rm P} + R_{\rm ECF} \right) C_{\rm P} \right]^{\frac{1}{\alpha_{\rm P}}}}.$$
 (9)

The open-source ZFit MATLAB script (Dellis, 2022) for fitting data to the equivalent circuit has been
used to implement the data fitting to this modified model.

236

237 2.3 Instrument calibration

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Usually, impedance measurement systems need calibration to correct for deviations in magnitude and phase. This was performed using a reference resistor of 1000 Ω whose magnitude and phase were measured with a 4294A impedance analyzer (Agilent). After calibration, a test impedance (Z_t) consisting of a 1-k Ω resistor in parallel with a series combination of another 1-k Ω and a 10-nF capacitor was measured with the proposed system. The results were compared with those obtained with the 4294A impedance analyzer.

245

246 2.4 Influence of electrode positioning on the measured impedance spectrum

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The current injecting and voltage pick-up electrodes were built from thin aluminum foil, forming narrow rectangles. 8-cm square samples were cut from fresh *spinacea oleracea L*. Electrodes I_1 and I_2 were placed at the upper face extremes, while electrodes E_1 and E_2 were placed at the same face, as shown in Figure 4. To decrease contact impedance, a small quantity of ECG gel was spread at the interface between the leaf and each electrode. Five distances *d* between electrodes were tested: 0,5 cm, 1 cm, 1,5 cm, 2 cm and 2,5 cm. The tests were repeated but with electrodes I_1 and I_2 placed at the lower face, while electrodes E_1 and E_2 were placed at the upper face, and the same distances (*d*) were tested again, totaling 10 different measurements.



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Figure 4. Electrodes were placed onto the square leaf samples at different distances (*d*) between E₁ and E₂.

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260

Three rectangular leaf samples of 5 cm × 10 cm, all from different *spinacea oleracea L.* plants, were left to dry spontaneously in a closed room with temperatures around 26° and relative humidity in the 70%- 80% range. The electrodes used were placed in the upper face, with d = 2 cm. Each leaf sample was placed on an electronic weighing scale (PCE-BS-300, PCE Instruments) with all the measuring electrodes in place. This weighing scale has an accuracy of ±0,03 g over a full scale of 300 g. The loss of weight of each leaf and its complex impedance spectrum were measured simultaneously at 6 different times during two days:

- Day1: initial measurement (t0), and after 3 h (t1) and 6 h (t2)
- Day 2: after 24 h (t3), 27 h (t4) and 30 h (t5)

270 RWC is a method based on weight loss during desiccation, with the underlying hypothesis that the 271 only mass loss is that due to evaporated water. Here, it was assumed that the progressive mass loss 272 during the free drying process was also due mostly to the water lost. Therefore, the approach used 273 here was to compare the trends of the equivalent-circuit parameter values against the leaf sample

weight during the drying process to identify those who are more sensible to water content loss.

- 275 **3.** Results and discussion
- 276

277 3.1 System calibration

278 The results of the magnitude and phase spectrum of Z_t before and after calibration are shown in 279 Figures 5a and 5b, respectively, and those obtained with the 4294A impedance analyzer are shown 280 in Figure 6. The frequency range was extended up to 10 MHz, although the intended frequency range 281 ends at 1 MHz. At 100 Hz, the reference measurements –obtained with the commercial impedance 282 analyzer– were 1001 Ω /-0,1 ° (magnitude/phase), while at 1 MHz they were 503 Ω /-0,1 °. Before the 283 calibration, the values obtained with the proposed system were 994 Ω /-0,3 ° at 100 Hz, and 502 Ω /-284 1,8 ° at 1 MHz. Those values changed after calibration to 1001 Ω /-0,2 ° at 100 Hz, and to 501 Ω /0 ° at 285 1 MHz. The minimum phase occurred at 11,7 kHz with the reference instrument, at 12,4 kHz with the 286 proposed system before calibration and at 12 kHz after calibration. These measurements confirm 287 that this low-cost system can provide reliable data in the frequency range from 100 Hz to 1 Mhz. 288 At frequencies larger than 1 MHz the largest deviations observed are in the phase spectra (5 ° larger 289 with the proposed system, at 10 MHz). This large phase error means that this system will not be able 290 to yield reliable data above 1 MHz.



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b) after calibration.



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Figure 6. Impedance spectrum (magnitude and phase) of *Z*_t obtained with the 4294A impedance analyzer.

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3.2 Electrode positioning

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300 The Cole-Cole plots for the measured impedances at different d values are displayed in Figure 7, 301 where the upper plot corresponds to I1 and I2 placed at the upper leaf face (UF in Table 1) and the 302 lower plot corresponds to the same electrodes placed on the lower leaf face (LF in Table 1). E1 and 303 E2 were always placed on the upper leaf face. The fitted electrical parameters of the double-shell 304 model are shown in Table 1. At d = 0.5 cm, the Cole-Cole plot exhibits two depressed arcs (Figure 8) 305 and the fitted values are generally in agreement with those obtained by (Zhang and Willison, 1993) 306 for cabbage leaves (*brassica oleracea*). Using CPEs instead of capacitors improved data fitting. 307 According to the equivalent circuit used, the high frequency arc (HFA) -to the left of the plot-

308 depends on the smaller structures formed by the vacuoles and its membrane – the tonoplast–, while 309 the cell wall and the cytoplasm originate the lower frequency arc (LFA) –to the right of the plot. The

310 resistive elements increased with *d* while the capacitive elements decreased. These changes are

mostly due to geometrical changes: losing water shrinks the cells and vacuoles, hence the membrane
surface decreases, and the conductive paths narrow.

313 The LFA is wider than the HFA at the smallest d, although the frequency range did not permit to see 314 a complete LFA because f_P is around 22 Hz. Because these values of the characteristic frequencies fall 315 outside the measurement range, they are considered unreliable. However, this feature also points to 316 the possibility of better discrimination of finer structures. When increasing d, the HFA becomes more 317 and more dominant and $f_{\rm P}$ and $f_{\rm TP}$ move close together, making their discrimination difficult. At larger 318 scales –larger *d* values – the α_P value increases. One interpretation of α_P in a CPE is that it a low value 319 indicates a broader dispersion in cell size distribution. Hence, an increase is a sign that the material 320 becomes homogeneous at a "large" scale. However, α_{TP} shows little change with d, possibly due to the 321 fact that vacuoles occupy a large fraction of the epidermis cells (89%) and of the mesophyll cells (79 322 %) (Winter et al., 1994), and do not show large variations if the turgor is uniform across the leaf, 323 independently of the cell sizes. A second observation is that the HFA arc is dominant at the larger d 324 values, hence if vacuole water is lost it will be better detected for those *d* values, and the LFA arc 325 would become less important.

In (Watanabe et al., 2017), *spinacea oleracea L.* leaves were thermally treated to investigate the capability of EIS to track structural changes –mainly geometrical– in the leaves after thermal and mechanical treatments. There, the replacement of CPEs instead of capacitors in the Hayden model – similar to this study– also led to good fitting results. The findings of that study also showed impedance changes correlated with the structural changes induces by the thermal treatments applied to the leaf samples.



Figure 7. Cole-Cole plots for the measured and fitted data at different distances *d*. The upper plot displays data measured

335 with I_1 and I_2 at the upper leaf face, while the lower plot is for the lower face.



336

Figure 8. Cole-Cole plot for the measured and fitted impedance data at d = 0.5 cm and with the injecting electrodes at both faces of the leaf.

339

340 The parameter values are not significantly different in both the UF and LF electrode setups (Table 1).
341 The implication is that the electrodes can be placed as deemed necessary to best fit the leaf
342 dimensions.

343

344 Data interpretation of the four-electrode impedance measurements depends also on geometrical
 345 considerations. The current density paths between injecting and pick-up electrodes determine the

346 sensitivity to the material impedance throughout the volume conductor and might create small 347 regions of zero impedance and even negative impedance that contribute to the overall measured 348 impedance (Grimnes and Martinsen, 2015). In (Zhang and Willison, 1993), the two-electrode 349 technique was used with electrodes inserted into the leaf, causing damage but lowering the contact 350 impedance, which can be large at low frequencies -masking the LFA-. But the two-electrode does 351 not have the problems of sensitivity of the four-electrode technique. In contrast, using four electrodes 352 largely attenuates the issues introduced by the electrode contact impedance, enabling noninvasive 353 measurements. The influence of the sensitivity in the measurements has not been considered in this 354 work, and is left for future research.

355

356 Table 1

Fitted parameters of the impedance data for all positions, as described in section 2.3. The parameters correspond to thedouble-shell equivalent circuit in Figure 3C. UF stands for the upper leaf face and LF for the lower leaf face.

d (cm) - I₁&I₂	R _{cw} (kΩ)	$R_{\rm CP}({ m k}\Omega)$	<i>С</i> _Р (µF)	α _P	R _{IV} (kΩ)	C_{TP} (nF)	$lpha_{ ext{TP}}$	f _P (Hz)	f _{TP} (kHz)
0,5-UF	24,88	5,40	3,288	0,47	2,35	9,99	0,79	21,9	25,5
1,0-UF	33,26	58,16	0,162	0,60	4,93	8,34	0,73	173,6	5,1
1,5-UF	85,23	116,11	0,025	0,67	6,71	7,79	0,70	443,2	3,1
2,0-UF	114,27	174,17	0,014	0,68	9,41	6,38	0,70	562,2	2,4
25-UF	164,43	262,42	0,007	0,71	12,84	4,55	0,70	588,6	2,1
0,5-LF	28,11	5,01	3,182	0,48	2,03	9,47	0,79	17,0	28,9
1,0-LF	40,67	86,44	0,137	0,64	4,19	6,67	0,74	92,0	3,7
1,5-LF	55,51	218,01	0,028	0,71	5,95	5,41	0,72	158,3	1,7
2,0-LF	82,24	328,64	0,006	0,75	10,78	2,97	0,75	492,6	1,6
2,5-LF	114,37	480,64	0,004	0,76	14,45	2,48	0,74	451,8	1,3

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365 Figures 9a, 9b and 9c show the Cole-Cole plots of the impedance spectrums of the three leave samples 366 during the drying process, at the times specified in section 2.5. All of them show one dominant arc 367 that corresponds to the HFA, as expected from the results obtained in section 3.2 with *d* values above 368 1,5 cm. For any of the leaves, the maxima of all arcs shows increases (an increase in the impedance 369 magnitude) as the leaf losses water, and its location moves toward the right (lower frequencies). 370 Because there is a dominant arc, a modified single-shell model (Figure 3A) was used. Table 2 shows 371 the fitted values for R_P , R_{ECF} , α_P , C_P , f_c , and the leaf mass *m*, where f_c is calculated according to (9), and 372 Table 3 shows the results from the linear regression analysis. From the latter, the parameters, $R_{\rm ECF}$ 373 and f_c show a consistent and strong linear correlation with *m* throughout all the leaf samples, but of 374 opposite sign. R_{ECF} showed R² values from 0,897 to 0,998 and increased with *m*. This behavior is 375 compatible with a loss of weigh due to water evaporation because the leaf becomes less electrically 376 conductive. On the other hand, this same behavior explains why f_c decreases –it is inversely 377 proportional to R_{ECF} - and keeps a consistent trend among the three leave samples.



1



Figure 9. Cole-Cole plots for the impedance data recorded from the three leaf samples (A, B and C) during the drying

386 process.

387

388 Table 2

389 Fitted parameters (R_{ECF} , R_P , C_P , α_P , and f_c) obtained from the three leaf samples (A, B, and C) while they were drying over

390 two days at times t0 to t5, and the corresponding measured leaf mass *m*.

Sample	Time	R_{ECF} (k Ω)	<i>R</i> _P (kΩ)	C _P (nF)	αP	f _c (kHz)	<i>m</i> (g)
A	t0	111,76	6,93	5,23	0,69	46,92	5,08
	t1	124,91	6,31	5,88	0,68	39,99	5,02
	t2	129,03	6,26	6,00	0,67	38,43	4,98
	t3	138,55	6,25	6,77	0,66	33,54	4,71
	t4	141,82	6,08	6,84	0,66	33,06	4,66
	t5	144,67	5,68	6,75	0,66	32,83	4,63
В	t0	169,30	7,23	4,98	0,67	34,94	4,49
	t1	179,66	6,82	5,13	0,67	31,98	4,45
	t2	188,75	6,63	5,08	0,67	32,41	4,42
	t3	229,94	7,44	3,67	0,68	30,20	4,28
	t4	235,88	6,58	3,38	0,69	30,18	4,25
	t5	245,33	7,45	3,37	0,68	30,28	4,23
	t0	157,94	7,25	8,59	0,65	24,74	4,52
с	t1	178,65	6,86	9,31	0,64	20,40	4,46
	t2	186,58	5,97	9,59	0,63	20,58	4,43
	t3	254,28	6,85	6,91	0,65	16,73	4,26
	t4	274,12	5,91	6,41	0,65	16,06	4,22
	t5	298,65	6,64	6,24	0,65	15,27	4,2

392

Table 3. Linear regression analysis results for the fitted parameters on the three leaf samples (A, B, and C). R² is the

394 regression coefficient and Trend indicates whether there is a clear increase or decrease of a specific parameter with *m*.

Sample	mvs.	R ²	Trend
	R ECF	0,897	decrease
	R _P	0,64	increase
A	CP	0,913	decrease
	αΡ	0,905	increase
	f _c	0,85	increase
	R ECF	0,999	decrease
	RP	0,061	-
В	CP	0,939	increase
	αΡ	0,786	decrease
	f _c	0,819	increase
	R _{ECF}	0,985	decrease
	R _P	0,167	-
С	CP	0,825	increase
	α _P	0,364	decrease
	f _c	0,934	increase

395

396 R_p is usually interpreted as the intracellular content of the plant cells, and was expected to also show 397 an increase due to water loss that was not confirmed experimentally. On the other hand, C_p was 398 expected to decrease with water loss due to cell shrinking but, again, was not confirmed in these 399 experiments. Possible damage on the cell membrane or in the vacuole membrane might provide an 400 explanation for this behavior, but in this work there were no microscope evidence and the hypothesis 401 cannot be further tested. Lastly, the small variations of α_p around a value of 0,6 might indicate that 402 the drying process does not increase or decrease the cell size dispersion and is not sensible enough403 to provide information about water loss.

404 Other studies have attempted to measure the effect of watering or water stress in plants, but most of 405 them do so at the plant stem, like (Comparini et al., 2020), (Bar-On and Shacham-Diamand, 2021) or 406 (Garlando et al., 2022), that has a completely different geometry: a thicker volume and higher water 407 content due to the plant sap. Therefore, the recorded impedances in those studies were lower as 408 compared to this work. The study by (Bar-On and Shacham-Diamand, 2021) also used the four-409 electrode technique and recorded the plant weight on a gravimetric platform. Observed increases of 410 the impedance corresponded to weight loss while watering events decreased it. In addition, an ad 411 hoc equivalent circuit was proposed where the resistive components increased with water 412 restriction and stress, which is compatible with the observed behavior here. Regarding the 413 capacitance and exponential of the CPEs, no data was shown, making it difficult to make further 414 comparisons. In (Comparini et al., 2020), two electrodes were inserted into the stem and only the 415 resistance values were reported during stress and watering events, showing the same trend in 416 resistance as in this study. Finally, (Garlando et al., 2022) studied the Bode plots of the impedance 417 and chose a frequency of around 10 kHz to record impedance changes, again confirming the same 418 behavior.

419

420 **4.** Conclusions

A cost-effective impedance analyzer for plant leaves was built. It was tested over a range frequency between 100 Hz and 1 MHz frequency range, demonstrating its capability to perform reliable fourelectrode impedance measurements. The electrodes used are very simple and cheap and can be used for noninvasive measurements. It has a simple architecture yet it yields a good quality impedance spectrum for plant leaves due to its fully-differential architecture. It is based on the Analog Discovery 426 2, a USB module that comprises the necessary power supplies, signal acquisition and signal427 generation circuitry.

428 The four-electrode method of impedance measurement avoids the adverse effects of the electrode 429 impedance on the impedance plots and enables a better observation of the plant impedance. 430 Increasing the distance between the measuring electrodes allows using a three-element -with a total 431 of four parameters – equivalent circuit, which is convenient for data analysis. From those parameters, 432 the resistance that models the extracellular fluid showed that it is consistently increasing with water 433 loss -as recorded by means of the weight decrease. In contrast to other studies that place the 434 electrodes into the plant stem, this study shows that water content can be measured on the leaves, 435 where the hydration deficit might manifest earlier than in the plant stem in a living plant. However, 436 such living-plant application needs of further development of suitable electrodes that do not disturb 437 the plant physiology.

438

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