



1 Time-lapse monitoring of root water uptake using electrical resistivity 2 tomography and Mise-à-la-Masse: a vineyard infiltration experiment

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10 **Abstract.** This paper presents a time-lapse application of electrical methods (Electrical Resistivity Tomography – ERT – and
11 Mise-à-la-Masse – MALM) for monitoring plant roots and their activity (root water uptake) during a controlled infiltration
12 experiment. The use of non-invasive geophysical monitoring is of increasing interest as these techniques provide time-lapse
13 imaging of processes that otherwise can only be measured at few specific spatial locations. The experiment here described was
14 conducted in a vineyard in Bordeaux (France) and was focused on the behaviour of two neighbouring grapevines. The joint
15 application of ERT and MALM has several advantages. While ERT in time-lapse mode is sensitive to changes in soil electrical
16 resistivity and thus to the factors controlling it (mainly soil water content, in this context), MALM uses DC current injected in
17 a tree stem to image where the plant-root system is in effective electrical contact with the soil at locations that are likely to be
18 the same where root water uptake (RWU) takes place. Thus ERT and MALM provide complementary information about the
19 root structure and activity. The experiment shows that the region of likely electrical current sources produced by MALM does
20 not change significantly during the infiltration study time in spite of the strong changes of electrical resistivity caused by
21 changes in soil water content. This fact, together with the evidence that current injection in the soil produces totally different
22 patterns, corroborates the idea that this application of MALM highlights the active root density in the soil. When considering
23 the electrical resistivity changes (as measured by ERT) inside the stationary volume of active roots delineated by MALM, the
24 overall tendency is towards a resistivity increase, which can be linked to a decrease in soil water content caused by root water
25 uptake. On the contrary, when considering the soil volume outside the MALM-derived root water uptake region, the electrical
26 resistivity tends to decrease as an effect of soil water content increase caused by the infiltration. The results are particularly
27 promising, and the method can be applied to a variety of scales including the laboratory scale where direct evidence of roots
28 structure and root water uptake can help corroborate the approach. Once fully validated, the joint use of MALM and ERT can
29 be used as a valuable tool to study the activity of roots under a wide variety of field conditions.

30



31 **1 Introduction**

32 The interaction between soil and biota is one of the main mechanisms controlling the exchange of mass and energy between
33 the Earth's terrestrial ecosystems and the atmosphere. Philip (1966) was the first to use the phrase "soil–plant–atmosphere
34 continuum" (SPAC) to conceptualize this interface in the framework of continuum physics. Even though more than five
35 decades have elapsed and many efforts have been expended (e.g., Maxwell et al., 2007; de Arellano et al., 2012; Anderegg et
36 al., 2013; Band et al., 2014), the current mechanistic understanding or modelling of SPAC of SPAC is still unsatisfactory (e.g.
37 Dirmeyer et al., 2006, 2014). This is not totally surprising, since soil-plant interactions are complex, exhibiting scale- and
38 species-dependence with high soil heterogeneity and plant growth plasticity. In this study, we focus on new methods to help
39 understand complex root-soil systems (the rhizosphere, e.g. York et al., 2016).

40 Roots contribute substantially to carbon sequestration. Roots are the connection between the soil, where water and nutrients
41 reside, to the other organs and tissues of the plant, where these resources are used. Hence roots provide a link in the pathway
42 for fluxes of soil water and other substances through the plant canopy to the atmosphere (e.g. Dawson and Stiegwolf, 2007).
43 These transpiration fluxes are responsible for the largest fraction of water leaving the soil in vegetated systems (Chahine,
44 1992). Root Water uptake (RWU) influences the water dynamics in the rhizosphere (Couvreur et al., 2012) and the partitioning
45 of net radiation into latent and sensible heat fluxes thereby impacting atmospheric boundary layer dynamics (Maxwell et al.,
46 2007; de Arellano et al., 2012). Yet, a number of issues remain when representing RWU in both hydrological and atmospheric
47 models. Dupuy et al. (2010) summarize the development of root growth models from its origins in the 1970s with simple
48 spatial models (Hackett and Rose, 1972; Gerwitz and Page, 1974) to the development of very complex plant architectural
49 models (Jourdan and Rey, 1997). Dupuy et al. (2010) advocate for a different approach, where roots systems are described as
50 "density" distributions. Attempts in this direction (Dupuy et al., 2005; Draye et al., 2010; Dupuy and Vignes, 2012) require
51 much less specific knowledge of the detailed mechanisms of meristem evolution, and yet are sufficient to describe the root
52 "functions" in the framework of continuum physics, i.e. the one endorsed by the SPAC concept. These models also lend
53 themselves more naturally to calibration against field evidence, as they focus on the "functioning" of roots, especially in terms
54 of RWU (e.g. Volpe et al., 2013, Manoli et al., 2014). However, calibration requires that suitable data are available in a form
55 comparable with the model to be calibrated. This is the main motivation behind the work presented herein.

56 A thorough understanding of root configuration in space and their evolution in time is impossible to achieve using only
57 traditional invasive methods: this is particularly true for root hairs, i.e. for the absorptive unicellular extensions of epidermal
58 cells of a root. These tiny, hair-like structures function as the major site of water and mineral uptake. Root hairs are extremely
59 delicate, turn over quickly, and are subject to desiccation and easily destroyed. For these reasons, direct investigation of their
60 in situ structure via excavation is practically impossible under field conditions.

61 The development of non-invasive or minimally invasive techniques are required to overcome the limitations of conventional
62 invasive characterization approaches. Non-invasive methods are based on physical measurements at the boundary of the
63 domain of interest, i.e. at the ground surface and, when possible, in shallow boreholes. Non-invasive methods provide spatially



64 extensive, high-resolution information that can also be supported by more traditional local and more invasive data such as soil
65 samples, TDR, lysimeters and rhizotron measurements.

66 In the investigation of roots and RWU the most widely used non-invasive technique is Electrical Resistivity Tomography (ERT
67 – e.g. Binley and Kemna, 2005). ERT measures soil electrical resistivity and, in time-lapse mode, resistivity changes over
68 time. Electrical resistivity values depend on soil type and its porosity, but also on state variables such as the saturation of
69 electrolyte (water) in the pores, and the concentration of solutes in the pore water (as described e.g. by the classical Archie's
70 law, 1942). Note, however, that other factors may play a role, such as clay content (Rhoades et al., 1976; Waxman and Smits,
71 1968) and temperature (e.g., Campbell et al., 1949). However, in general, it is possible to estimate water content changes from
72 changes in electrical resistivity over time (and space) provided that pore water salinity does not vary dramatically. While ERT
73 has been attempted for quantifying root biomass on herbaceous plants (e.g. Amato et al., 2009), the main use of this technique
74 in this context aims at identifying changes in soil water content in space and evolution in time (e.g., Michot et al., 2003, 2016;
75 Srayeddin and Doussan, 2009; Garré et al., 2011; Cassiani et al., 2012, Brillante et al. 2015). With specific reference to RWU,
76 Cassiani et al. (2015, 2016), Consoli et al. (2017) and Vanella et al. (2018) used time-lapse ERT with 3D cross-hole
77 configurations to monitor changes in soil electrical resistivity caused by irrigation and RWU for different crops (apple and
78 citrus trees). It should also be noted that RWU and the release of different exudates by fine roots modify soil water content and
79 resistivity at several temporal scales (York et al., 2016).

80 On the other hand, evidence suggests that roots themselves may produce signals in ERT surveys (Amato et al., 2008; Werban
81 et al., 2008); however, these signals are often difficult to separate from soil heterogeneities and soil water content variations
82 in space. Nevertheless, in most cases, the ranges of electrical resistivity of soil and roots overlap, and while the amplitude of
83 contrasts varies according to the soil resistivity and tree species (e.g. Mary et al., 2016), the direct identification of root systems
84 using ERT is often impractical.

85 Other electrical signals may contribute to the detection of roots and to the characterization of their activities. For instance, self-
86 potential (SP) signals can be associated with plant activities: water uptake generates a water circulation and a mineral
87 segregation at the soil–roots interface that induce ionic concentration gradients which in turn generate voltages of the order of
88 a few mV (Gibert et al., 2006). However, such SP sources are generally too low to be detectable in normally noisy environment.
89 Induced Polarization (e.g. Kemna et al., 2012) is also a promising approach in root monitoring. This is consistent with the fact
90 that root systems are commonly modelled as electrical circuits composed of resistance R and capacitance C (e.g. Dalton, 1995
91 and similar models). Recently, Mary et al. (2017) considered polarization from soil to root tissues, as well as the polarization
92 processes along and around roots, to explain the phase shift (between injected current and voltage response) observed for
93 different soil water content. Weigand and Kemna (2017, 2019) demonstrated that multi-frequency electrical impedance
94 tomography is capable of imaging root systems extent.

95 Recently, the Mise-A-La-Masse (MALM) method has been proposed for plant root mapping. MALM is a classical electrical
96 method (Parasnis, 1967) originally developed for mining exploration, but also used more recently e.g. in the context of landfill
97 characterization (De Carlo et al., 2013) as well as conductive tracer test monitoring (Osiensky, 1997; Perri et al., 2018). In



98 MALM, an electrical current is injected into a conductive body with a return current electrode far away (“at infinity”), and the
99 resulting voltage is measured at the ground surface or in boreholes, again with a reference electrode at infinity: the shape of
100 voltage contour lines is informative about the extent and orientation of the conductive body. This idea can be applied to the
101 plant stem and roots system, considering that electrical current can be transmitted through the xylem and phloem (on either
102 side of the cambium), where sap flow takes place. The main assumption is that fine root connections and mycorrhiza at the
103 contact between roots and soil convey the injected current into the soil where this contact is efficient, thus appearing as a
104 distribution of current sources in the ground. The location of these sources should correspond to the locations of active contacts
105 between roots and soil, and could be identified starting from the measured voltage distribution at the ground surface or in
106 boreholes. This approach has been recently test by Mary et al. (2018) on vine trees, showing that current injection in the stem
107 and in the soil just next to the stem produces very different voltage patterns, thus confirming that the stem-roots system conveys
108 current differently from a direct injection in the ground.

109 In this study we present the results of an infiltration experiment conducted in a Bordeaux vineyard (France). The experiment
110 was monitored (also) using time-lapse 3D ERT and time-lapse MALM measurements, the latter performed by injecting current
111 in the vine trees stems. This study had the following goals:

- 112 (a) define a non-invasive investigation protocol capable of “imaging” the root activity as well as the distribution of active
113 roots, at least in terms of their continuum description mentioned above;
- 114 (b) Integrate the geophysical results with mass fluxes measurements in/out of the soil-plant continuum system.

115

116 **2 Methodology**

117 **2.1 Site description**

118 The study was conducted in a commercial vineyard (Chateau La Louviere, Bordeaux) in the Pessac Leognan Appellation of
119 France (long 44°44’15’’N, lat 0°34’45’’W). The climate of the region is oceanic with a mean annual air temperature of 13.7
120 °C and about 800 mm annual precipitation. Grapevine trees are planted at 1 m distance along the rows, and the rows are spaced
121 about 1.5 m. We focused our interest on two neighbouring plants.

122 The vineyard is not irrigated. The soil is sandy down to 1 m depth with sandy clay below, down to 1.75 m, and calcareous at
123 depth. Due to its high porosity, the sandy layer has a relatively poor water retention capacity. Nevertheless, the water supply
124 of the vine plant is not a limiting factor - refer to Mary et al. (2018) for more details about the plants and soil type.

125 We concentrated our monitoring on only two neighbouring grapevines (Figure 1), which differed in age and size: plant A was
126 smaller and younger, plant B was considerably larger and older.

127



128 2.2 Meteorological measurements and irrigation schedule

129 Hourly meteorological data were acquired by an automatic weather station located about 300 m from the plot and managed by
130 DEMETER (Agrometeorological Service - www.meteo-agriculture.eu/qui-sommes-nous/lhistoire-de-demeter). These
131 micrometeorological data were valuable to estimate the initial soil conditions and the changes in time (Figure 2). Potential
132 evapotranspiration (ETP) was computed according to the Penman-Monteith formula accounting for the incoming short-wave
133 solar radiation, air temperature, air humidity, wind speed and rainfall measured by the station. Prior to June 19, 2017, date of
134 the first field data acquisition, little precipitation was recorded for 5 days (only 2.5mm on June 13) and only 18mm cumulative
135 precipitation was recorded during the entire month of June 2017. The mean air temperature was very high (35°C under a well-
136 ventilated shelter). Consequently, the plants were probably suffering from water deficit at the time of the experiment. Thus, at
137 the start of the experiment, we assumed that the soil water content (SWC) around the plants was probably below field capacity.
138 As shown in Figure 2, the evapotranspiration rate was about 5.6 mm/day.

139 The controlled infiltration experiment was conducted using a sprinkler installed between the two monitored plants, placed at
140 an elevation of 1.4m, in order to apply irrigation water as uniformly as possible. The irrigation started on June 19, 2017 at
141 13h00 and ended two hours later (15h00) for a total of 260 liters (104 l/h). Runoff was observed due to topography and probably
142 induced more water supply for plant A that is located downhill. The irrigation water had an electrical conductivity of 720µS/cm
143 at 15°C.

144 2.3 ERT and MALM data acquisition

145 We carried out a time-lapse microscale ERT acquisition, based on custom-made ERT boreholes (six of them, each with 12
146 electrodes), plus surface electrodes. The six boreholes were placed to form two equal rectangles at the ground surface. Each
147 rectangle size was 1 m by 1.2m respectively in the row and inter-row line directions, with a vine tree placed at the centre of
148 each rectangle. The boreholes were installed in June 2015 and a good electrical contact with soil was already achieved at the
149 time of installation. The topmost electrode in each hole was 0.1 m below ground, with vertical electrode spacing along each
150 borehole equal to 0.1 m. In each rectangle, 24 surface electrodes surrounded the plant stem arranged in a five by five regular
151 mesh (with one skipped electrode near the stem). We conducted the acquisitions on each rectangle independently. Each
152 acquisition was therefore performed using 72 electrodes (24 surface and 48 electrodes in 4 boreholes) using an IRIS Syscal
153 Pro resistivity meter. For all measurements we used a skip 2 dipole-dipole acquisition (i.e., a configuration where the current
154 dipoles and potential dipoles are three times larger than the minimal electrode spacing).

155 In addition to acquiring ERT data, we also acquired MALM data. MALM acquisition was logistically the same as ERT and
156 was supported by the same device, but used a pole-pole scheme (with two remote electrodes). Borehole and surface electrodes
157 composing the measurement setup were used as potential electrodes, while current electrode C1 was planted directly into the
158 stem, 10 cm from the soil surface, with an insertion depth of about 2 cm, in order to inject current directly into the cambium
159 layer. The two remote electrodes C2 (for current) and P2 (for voltage) were placed approximately at 30m distance from the



160 plot, in opposite directions. Note that for MALM (unlike than for ERT), one corner surface electrode was put near the stem in
161 order to refine the information at the centre of each rectangle.
162 Each MALM acquisition was accompanied by a companion MALM acquisition where the current electrode C1 was placed
163 directly in the soil next to the stem rather than in the stem itself. In this way the effect of the plant stem-root system in conveying
164 current can be evidenced directly comparing the resulting voltage patterns resulting from the two MALM configurations.
165 For both ERT and MALM, we acquired both direct and reciprocal configurations (that swap current and voltage electrode
166 pairs), in order to assess the reciprocal error as an estimate of measurement error (see e.g. Cassiani et al., 2006). We adopted
167 a time-lapse approach, conducting repeated ERT and MALM acquisitions over time in order to assess the evolution of the
168 system's dynamics under changing moisture conditions associated with the infiltration experiment. We conducted repeated
169 measurements starting on 19 June 2017 at 10:20 LT, and ending the next day at about 17:00 LT. The schedule of the
170 acquisitions and the irrigation times is reported in Table 1.

171 **2.4 Data analysis and processing**

172 **2.4.1 Micro-ERT time lapse analysis**

173 The inversion of ERT data was conducted using the classical Occam's approach (Binley and Kemna, 2005). We conducted
174 both absolute inversions and time-lapse resistivity inversions, as done in other papers (e.g. Cassiani et al., 2015, 2016). We
175 used for inversion only the data that pass the 10% reciprocal error criterion at all measurement times. We inverted the data
176 using the R3t code (Binley, 2019) adopting a 3-D mesh with very fine discretization between the boreholes, while larger
177 elements were used for the outer zone.

179 **2.4.2 MALM modelling and source inversion**

180 We measure the voltage V at N points, corresponding to the N electrodes locations, x_1, x_2, \dots, x_N . Voltage depends on the
181 density of current sources C according to Poisson's equation:

$$182 \quad \nabla \cdot (\sigma \nabla V) = C \quad (1)$$

183 where σ is the conductivity of the medium, here assumed to be defined by the conductivity distribution obtained from ERT
184 data inversion. The main idea behind the source inversion is to identify the distribution of current sources $C(x,y,z)$ – in practice
185 located at the mesh nodes $C=[C_1, C_2, \dots, C_{N_s}]$ – that produce the measured voltage V distribution in space. In general, given a
186 distribution of current sources it is relatively straightforward to calculate the resulting V field, once $\sigma(x,y,z)$ is known from
187 ERT inversion. Vice versa, the identification of $C(x,y,z)$ distribution given $V(x,y,z)$ and $\sigma(x,y,z)$ is an ill-posed problem, that
188 requires regularization and/or a priori assumptions in order to deliver stable results. Different approaches are possible – for a
189 detailed analysis in this context see Mary et al. (2018). In this paper we have used the simplest approach, i.e. we assumed that
190 one single current source was responsible for the entire voltage distribution. For each candidate location the sum of squares



191 between computed and measured voltages was used as an index of “likelihood” of that location as a possible MALM current
192 source in the ground. Mary et al. (2018) introduced a simple index that can be mapped in the three-dimensional soil space and
193 that measures the “likelihood” that a specific location is the (single) current source generating the observed voltage field. This
194 index ($F1$) is defined as:

$$195 \quad F1(D_m, D_{f,i}) = \|D_m - D_{f,i}\|_2 \quad (2)$$

196 Where D_m is the measured voltage, and $D_{f,i}$ is the modelled voltage corresponding to a single source injecting the entire known
197 injected current at the i -th node in the mesh. The forward modelling producing the $D_{f,i}$ values is based on the direct solution of
198 the DC current flow in a heterogeneous medium, such as implement in the R3t Finite Element code (Binley, 2019). Thus the
199 $F1$ inversion accounts naturally for the heterogeneous electrical resistivity of the 3D soil volume, also in its evolution over
200 time (e.g. as an effect of irrigation and RWU).

201 While more advanced attempts could be made (such as the $F2$ approach also described by Mary et al., 2018) the simple $F1$
202 approach is capable of imaging the likely location of current sources in the ground, that in turn represent - according to our
203 key assumptions – the locations where roots have an active contact with the soil.

204 **3 Results and discussion**

205 **3.1 Background and irrigation time steps of ERT measured data**

206 The soil electrical conductivity during the period prior to the infiltration (see ERT results in Figure 2a and 2b, respectively for
207 Plants A and B) ranged from 50 to 200 Ωm , with a median value around 100 Ωm , a range that is reasonable for a dry sandy
208 soil. For plant A, i.e. the smaller plant, the highest resistivity values were distributed at about 0.5 m depth (Figure 3a), For the
209 larger plant B (Figure 3c), the positive resistivity anomalies are more diffused and less intense (150 Ωm) compared to plant
210 A, which reach larger depths. The background time (T_0) for both plants revealed a low resistive layer ranging from 0 to 0.35
211 m depth for plant A and to 0.25m for plant B. The first time step (T_1) was collected during the irrigation, at 2h for plant A and
212 at 30 minutes for plant B after the beginning of the irrigation. Figure 3b and Figure 3d show the resistivity distribution during
213 irrigation (at time step T_1). The input of low resistivity water (15 Ωm , measured in laboratory) caused a homogeneous drop of
214 the resistivity values that make the two images around plant A and plant B very similar to each other, which is an indirect
215 evidence that water infiltrated similarly in both areas (that are next to each other) with no difference in soil hydraulic properties.
216 The very small-scale anomalies observed at the soil surface are likely to be caused by heterogeneous direct evaporation patterns
217 or different soil compaction. More interesting are the resistive anomalies at intermediate depths. As observed in other case
218 studies (e.g. Cassiani et al., 2015, 2016, Consoli et al., 2017; Vanella et al. 2018), these are likely to be linked to soil saturation
219 decrease caused by RWU, particularly in consideration of its intensity during this time of the year (June) for non-irrigated
220 crops. Of course, we cannot fully exclude that higher resistivity is also related to woody roots presence, especially when they
221 are dense. Besides, roots could also have induced soil swelling creating voids acting like resistive heterogeneities. Consistently



222 to the age and size of plants, plant A that is smaller and younger is associated to an estimated root system that is shallower and
223 more concentrated (Figure 3a) while plant B seems to have a root system deeper and larger (Figure 3c).

224 3.2 Background and irrigation time steps of MALM measured data

225 Figure 4 shows the raw results of MALM acquisition on plant A, during background and irrigation, for both soil and stem
226 injection configurations. Note that voltages are normalized against the corresponding injected current. For both surface and
227 borehole electrodes the normalized voltage distribution can be compared against the one expected from the solution for a point
228 injection of current I at the surface of a homogeneous soil of resistivity ρ :

$$229 \quad V = \frac{I\rho}{2\pi r} \quad (3)$$

230 where r is the distance between the (surface) injection point and the point where voltage V is computed (see Figure 4e for a
231 comparison). In all cases, both for surface and borehole electrodes, and both for stem and soil current injection, the voltage
232 patterns are deformed with respect to the solution of Eq. (3) for a homogeneous soil. Some pieces of evidence are apparent
233 from the raw data already:

- 234 a. In all cases, the pattern of surface and subsurface voltage is asymmetric with respect to the injection point (in the stem
235 or close to it, in the soil) and thus different from the predictions of Eq. (3); this indicates that current pathways are
236 controlled by the soil heterogeneous structure: note that at all times there is a clear indication that a conductive
237 pathway extends from the plant to the right-upper corner of the image (this would be the classical use of MALM –
238 identifying the shape of conductive bodies underground). Note that spatial variations of voltage between boreholes
239 are consistent with surface observations i.e. the maximum voltage was measured on the borehole 4 located in the top
240 right corner of the plot;
- 241 b. The voltage patterns in the case of stem injection are clearly different from the corresponding ones obtained from soil
242 injection. In particular, injecting in the soil directly produces a stronger voltage signal both at the surface and in the
243 boreholes than the corresponding voltage in the case of stem injection: this difference clearly points towards the fact
244 that the plant-roots system must convey the current in a different way than the soil alone; tentatively the observed
245 voltage features would indicate a deeper current injection in the case of stem injection. Looking at the qualitative
246 differences between soil and stem injection in the borehole electrode data, the impact is very small at depths larger
247 than 0.6m;
- 248 c. For both soil and stem injection, local anomalies observed in the background image are either removed or smoothed
249 during the irrigation steps. The effect is equally pronounced in soil and stem injection, showing that this is caused
250 essentially by the change in resistivity induced by the change in soil water content (see Figure 3Figure 4).

251 Similar features are observed for plant B (results not shown for brevity).



252 Figure 4 shows the time evolution of normalized voltage values for both stem and soil injection for plant A. As apparent from
253 the raw data in Figure 4, during irrigation all electrodes recorded a decrease of the normalized voltage. However, a consistent
254 and quantitative interpretation is not straightforward by a visual inspection of the raw data (Figure 4, or similar data for Plant
255 B).

256 3.3 Inversion of virtual current sources to estimate roots extents

257 Figure 6 shows the iso-surfaces of fitness index FI (Eq. 2) for the background (pre-irrigation) conditions of both plants A and
258 B, and for current injection in the soil and in the stem at all-time steps listed in Table 1. In all cases, Figure 6 shows the iso-
259 surface corresponding to the value $FI=7V$ corresponding to the 25% percentile (value selected after analysing the evolution
260 of the curve of sorted misfit FI), thus the images provide comparable information for all cases. In particular, the inversion
261 procedure highlights the remarkable difference, for both plants A and B, between the injection in the stem and in the soil.
262 Current injection in the soil produces a voltage distribution that, albeit corresponding to a heterogeneous resistivity distribution
263 and thus different from the predictions of a simpler model such as Eq. (3), collapses effectively to one point, i.e. the point
264 where current was effectively injected in the ground. On the contrary, when current is injected in the stem, the region of
265 possible source locations in the ground is much wider, and depicts a volume that is likely to correspond to the contact points
266 between roots and soil, i.e. the volume where roots have an active role in the soil especially in terms of RWU. While this latter
267 interpretation remains somewhat speculative, at least in the present experimental context, nevertheless the different results
268 between soil and stem injection are absolutely apparent and can only find an explanation in the role of roots and their spatial
269 structure.

270 The most interesting feature shown by Figure 6 is that the likely source volumes do not change with time during irrigation.
271 Note that this inversion makes use of the changing electrical resistivity distributions caused by infiltrating water (see Figure
272 3) thus the result is not obvious, and indicates an underlying mechanism that is likely to be linked to the permanence of the
273 roots structure over such a short time lapse.

274 3.4 Electrical resistivity variations inside and outside the likely active roots zone

275 As the changes in the estimated extent of the root zone are only minor (Figure 6), it makes sense to evaluate the changes, as
276 an effect of irrigation, in electrical resistivity within such stable estimated root zone. Figure 7 shows the ER variations in the
277 zones inside and outside this estimated active root zone. It is apparent how irrigation causes a general decrease of electrical
278 resistivity for both plants A and B, and in both inner and outer regions. Note that even though the regions are different for the
279 two plants, the behaviour is similar. Then at the end of irrigation we observe, for both plants, that resistivity continues to
280 decrease outside the root active region, while it increases slightly inside. This behaviour is consistent with the fact that inside
281 the region we expect that RWU progressively dries the soil, while outside this region resistivity continues to decrease (overall)



282 as an effect (probably) of water redistribution in the unsaturated soil. It must be noted that the relationships between resistivity
283 and soil moisture content is non-linear (e.g. Archie, 1942; Waxman and Smits, 1968; Brovelli and Cassiani, 2011).

284 **4 Conclusions**

285 This study presents an approach to define the extent of active roots distribution using non-invasive investigations, and thus
286 particularly suitable to be applied under real field conditions. We applied a mix of ERT and MALM techniques, using the
287 same electrode and surface electrode distribution. The power of the approach lies in the complementary capabilities of the two
288 techniques in providing information concerning the root structure and activity:

- 289 (a) ERT provides 3D high resolution images of electrical resistivity distribution in the subsoil housing the root system.
290 Fast acquisition allows the measurement of resistivity changes over time, which in turn can be linked to changes in
291 SWC. This can be caused e.g. by water infiltration, or by RWU: in the latter case, negative SWC changes mapped
292 through resistivity changes can be used to map the regions where roots exert an active suction and reduce SWC.
293 However, water redistribution in the soil also plays a role in terms of resistivity changes. Thus some additional
294 independent information about the location of active roots in the soil may help: this is the first coupling between ERT
295 and MALM.
- 296 (b) MALM, and particularly its double application of current injection in the stem and in the soil next to it, uses electrical
297 measurements in a totally different manner: here the plant-root system itself acts as a conductor, and the goal is to
298 use the retrieved voltage distribution to infer where the current injected in the stem actually is conveyed into the soil:
299 these locations are potentially the same locations where roots interact with the soil in terms of RWU. However, in
300 order to try and locate the position of these points, it is necessary to know the soil electrical resistivity distribution at
301 the time of measurements. This can be provided by (time-lapse) ERT measurements: this is the second coupling
302 between ERT and MALM.

303 The approach has been successfully tested in a vineyard during an irrigation experiment. The survey was carried out during a
304 sunny summer season in a generally non-irrigated vineyard of the Bordeaux Region. The site is composed of sandy-loamy
305 soil, thus there is a high infiltration rate during the experiment, and this would make it more difficult to distinguish RWU zones
306 from infiltration zones as done for instance by Cassiani et al. (2015) using time-lapse ERT alone.

307 The key additional information is provided by MALM, and particularly by the simple inversion seeking for a single current
308 injection point that we have adopted. This directly incorporates the ERT information in terms of changing electrical resistivity
309 distribution in space including its evolution in time.

310 On one hand, this gives clear evidence that injecting current in the soil and in the stem produces different inversions. The soil
311 injection leads practically to identifying the true single electrode location. The stem injection helps identify a 3D region of
312 likely distributed current injection locations, thus defining a region in the subsoil where RWU is likely to take place.



313 The latter result is particularly useful, in perspective: when computing the time-lapse changes of electrical resistivity inside
314 and outside this tentative RWU region during irrigation we clearly see that while inside resistivity increases (as an effect of
315 RWU, as irrigation is still ongoing), outside resistivity decreases. Thus our assumption that the region identified by MALM
316 inversion (albeit very rough) corresponds to the RWU region is corroborated indirectly also by this evidence.
317 The presented approach can be easily replicated under a variety of conditions, as DC electrical methods such as ERT and
318 MALM do not possess a spatial scaling per se, but their resolution depends on electrode spacing as well as on other factors
319 that are difficult to assess a priori, such as resistivity contrasts and signal to noise ratio. Thus similar experiments can also be
320 used in the laboratory, where more direct evidence of root distribution can be used to further validate the method.

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454 **Table 1: schedule of the acquisitions and the irrigation times; Plant A and B are measured consecutively and consist each time of**
 455 **three measurements: ERT, MALM stem and MALM soil.**

Acquisition no.	Plant	Starting time (LT)	Ending time (LT)	Irrigation	Date
0 (background)	A	10:20	11:00		
	B	12:20	13:00		
1 (Irrigation)	A	15:00	15:30	13h00 to 15h30, 104lh-1 For both plants	Day 1 (19 June 2017)
	B	13:30	14:00		
2	A	17:00	17:30		
	B	18:00	18.45		
3	A	10:30	11:00		
	B	9:30	10:00		
4	A	14:00	14:30		Day 2
	B	15:00	15:30		
5	A	18:00	18:30		
	B	17:00	17:30		

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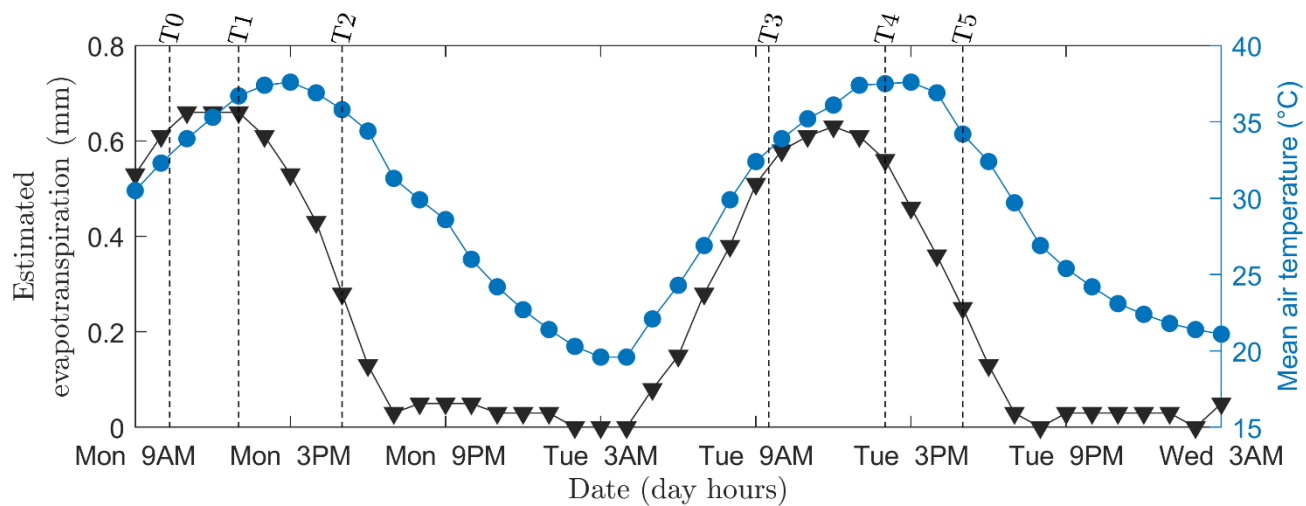


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Figure 1: picture of the field site in May 2017 (a) wired plants investigated (b) and grape status during the experiment in June 2017 (c)



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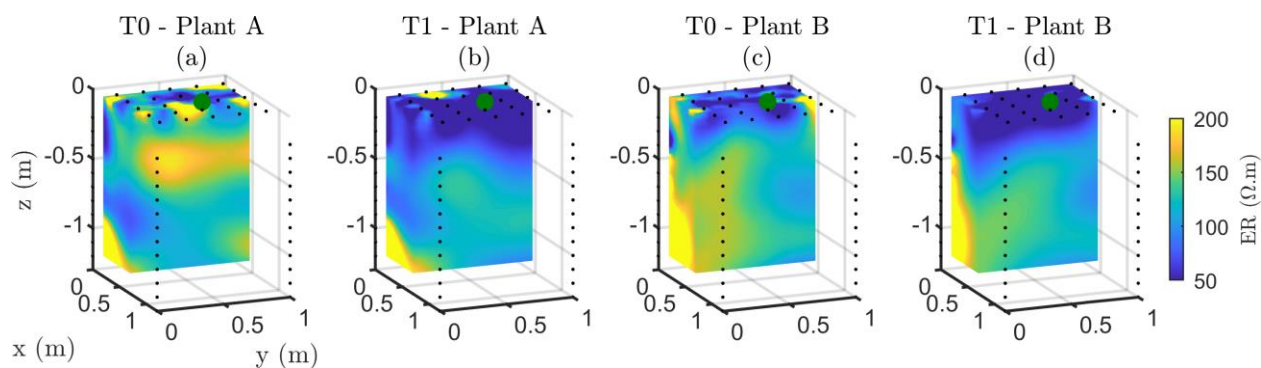


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Figure 2: variation of temperature (blue line) and estimated evapotranspiration (black line) derived from a nearby meteorological station, during the 2-day irrigation experiment. The dashed vertical lines indicate acquisition times (see Table 1).



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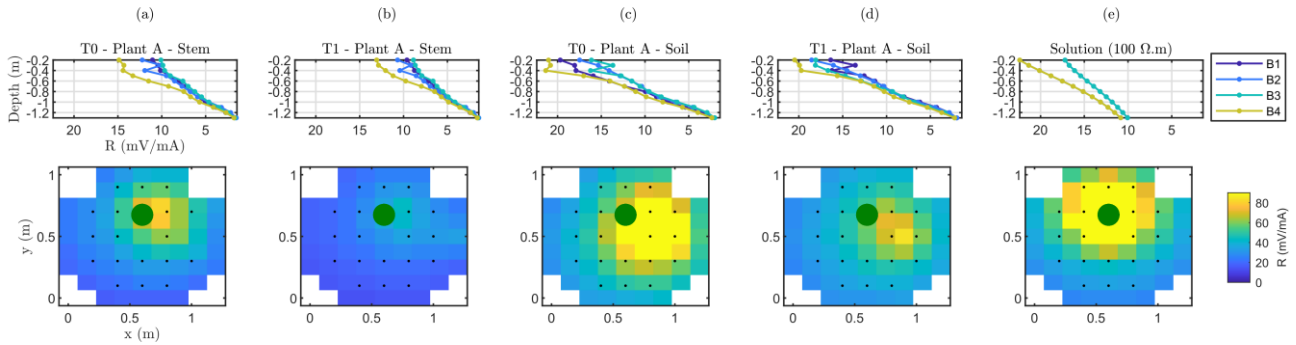
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Figure 3: ERT results showing the 3D variations of electrical resistivity (in Ωm) for the initial state background T0 (a,c) and during irrigation T1 (b,d), for both plants A and B. The green point shows the positions of the plant stem.



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473 **Figure 4: plant A, MALM results showing variations in surface (horizontal plan) of resistance R (in mV/mA) for the initial state**
474 **background T0 (a,c) and irrigation T1 (b,d) time steps. Comparison between the stem injection (a,b) and soil injection (c,d). The**
475 **black points show the surface electrodes location. The green point shows the positions of the plant stem. Data are filtered using a**
476 **threshold on reciprocal acquisition of 20%. (e) shows the solution using eq. (2) for a homogeneous soil of 100 Ohm.m; The resistance**
477 **between boreholes B1/B3 and B2/B4 are identical and cannot be distinguished graphically in the case of (e).**

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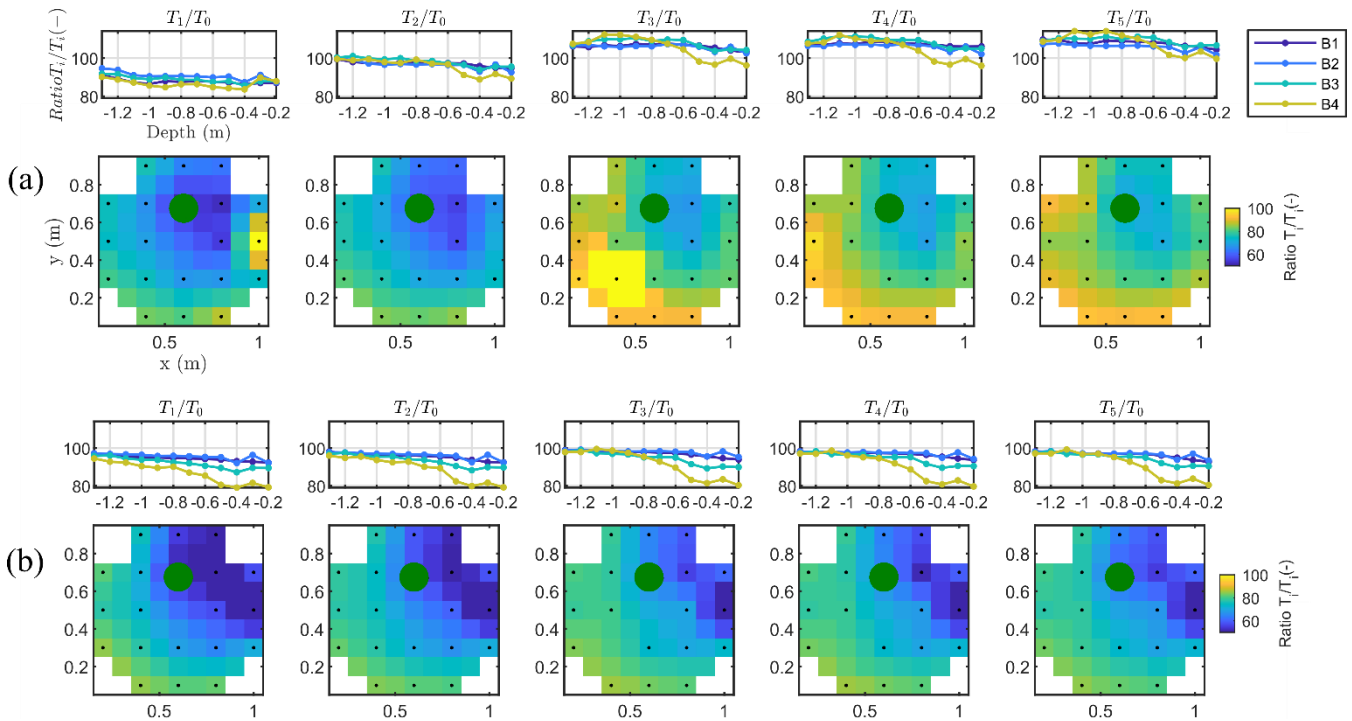
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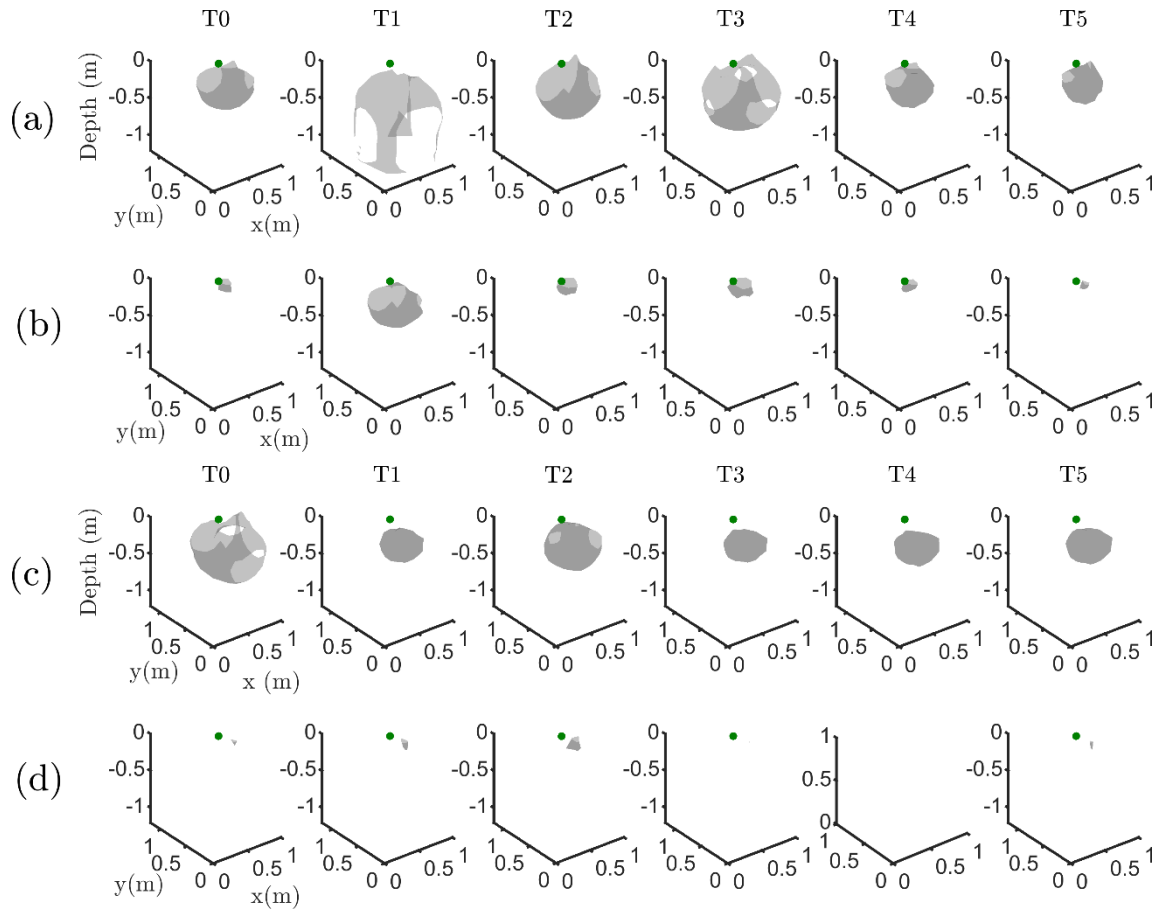
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484 **Figure 5: normalized voltage ratios (unitless) between time dataset T_i and T_0 (T_i/T_0), measured for plant B. Above each plan view**
485 **plot are the normalized voltage data measured using the boreholes electrodes, while map plots show ratios of normalized voltages**
486 **measured with surface electrodes. Comparison between injection within (a) the stem (lines 1 and 2) and (b) in the soil (lines 3 and**
487 **4). The green point shows the positions of the plant stem.**

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491 **Figure 6: isosurface minimizing the FI function for plant B (a,b) and plant A (c,d); during stem injection (a,c), during soil injection**
492 **(b,d); Columns represent the six times steps from T0 to T5. Green dot shows plant stem position. Threshold is defined by the**
493 **percentile 25% of the normalised FI (value selected according to the evolution of the curve of sorted misfit FI and calculated for the**
494 **tree injection at T0 and kept constant for all the time steps).**

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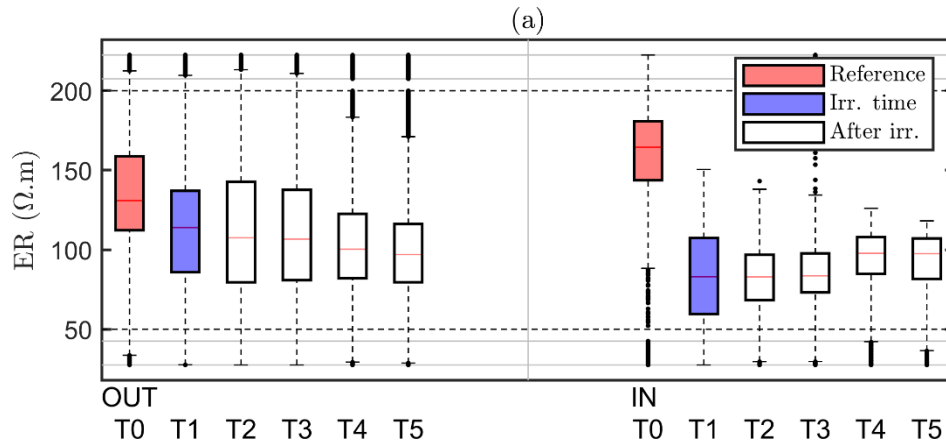
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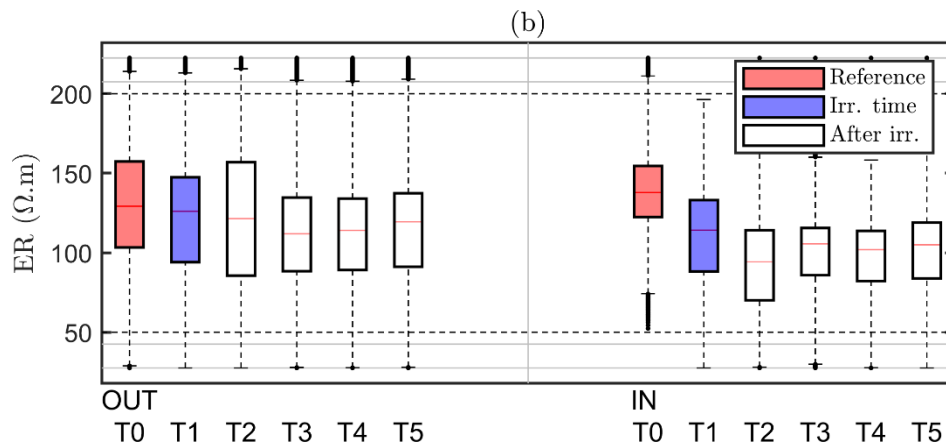
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509 **Figure 7: boxplot distribution of ER time variations observed on the plant A (top) and plant B (bottom), outside (OUT) of the region**
510 **defined by the F1 best fit sources (see Fig. 6a-T0). The central mark indicates the median, the bottom and top edges of the box**
511 **indicated the 25th and 75th percentiles of ER data, respectively. The whiskers extend to the most extreme data points not considered**
512 **outliers.**