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6	Monitoring and modelling of soil-plant interactions:
7	the joint use of ERT, sap flow and Eddy Covariance data
8	to characterize the volume of an orange tree root zone.
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23 Abstract

Mass and energy exchanges between soil, plants and atmosphere control a number of key 24 environmental processes involving hydrology, biota and climate. The understanding of these 25 exchanges also play a critical role for practical purposes e.g. in precision agriculture. In this paper 26 we present a methodology based on coupling innovative data collection and models in order to 27 obtain quantitative estimates of the key parameters of such complex flow system. In particular we 28 propose the use of hydro-geophysical monitoring via "time-lapse" Electrical Resistivity 29 Tomography (ERT) in conjunction with measurements of plant transpiration via sap flow and 30 31 evapotranspiration from Eddy Covariance (EC). This abundance of data is fed to spatially distributed soil models in order to characterize the distribution of active roots. We conducted 32 experiments in an orange orchard in Eastern Sicily (Italy), characterized by the typical 33 34 Mediterranean semi-arid climate. The subsoil dynamics, particularly influenced by irrigation and root uptake, were characterized mainly by the ERT setup, consisting of 48 buried electrodes on 4 35 36 instrumented micro boreholes (about 1.2 m deep) placed at the corners of a square (about 1.3 m in side) surrounding the orange tree, plus 24 mini-electrodes on the surface spaced 0.1 m on a square 37 grid. During the monitoring, we collected repeated ERT and TDR soil moisture measurements, soil 38 39 water sampling, sap flow measurements from the orange tree and EC data. We conducted a laboratory calibration of the soil electrical properties as a function of moisture content and pore 40 water electrical conductivity. Irrigation, precipitation, sap flow and ET data are available allowing 41 42 knowledge of the system's long term forcing conditions on the system. This information was used to calibrate a 1D Richards' equation model representing the dynamics of the volume monitored via 43 3D ERT. Information on the soil hydraulic properties was collected from laboratory and field 44 45 experiments. The successful results of the calibrated modelling exercise allow the quantification of the soil volume interested by root water uptake. This volume is much smaller (with a surface area 46

less than 2 square meters, and about 40 cm thickness) than expected and assumed in the design of
classical drip irrigation schemes that prove to be losing at least half of the irrigated water which is
not uptaken by the plants.

50 **Keywords**: Hydro-geophysics; Soil moisture; ERT, Eddy Covariance; Sap Flow; Root-zone.

51

52 1. INTRODUCTION

53 The system made of soil, vegetation and the adjacent atmosphere is characterized by complex patterns, structures, and processes that act on a wide range of time and space scales. While the 54 exchange of energy and water is continuous between compartments, the pertinent fluxes are 55 strongly heterogeneous and variable in space and time and this makes their quantification 56 particularly challenging. Plants are known to impact the terrestrial water cycle and underground 57 58 water dynamics through evapo-transpiration (ET) and root water uptake (RWU). The mechanisms of water flow in the root zone are controlled by soil physics, plant physiology and meteorological 59 60 factors (Green et al., 2003a). The translation of plant water use strategies into physically-based 61 models of root water uptake is a crucial issue in eco-hydrology and has fundamental consequence 62 in the understanding and modelling of atmospheric as well as soil processes. Still, no consensus 63 exists on the modelling of this process (Feddes et al., 2001; Raats, 2007). From a conceptual point 64 of view, two main approaches exist today, which differ in the way of predicting the volumetric rate of RWU. 65

A first approach expresses water transport in plants as a chain process based on a resistance law. Coupled with a three-dimensional soil water flow model, this approach leads to fairly accurate RWU models at the plant scale (Doussan et al., 2006; Schneider et al., 2010), also under water stress conditions. The limitations of these models are the cost of characterizing parameters, such as root system architecture and conductance to water flow, and their computational demand. A second approach, mostly used in soil-vegetation-atmosphere transfer models, relies on "macroscopic parameters" and predicts RWU as a product of the potential transpiration rate, a spatially distributed root parameter (e.g. relative root length density), and a stress function, depending on soil water potential and a compensatory RWU function (Jarvis, 1989). The major drawback of this approach is the necessity to calibrate the macroscopic parameters, which introduces substantial uncertainties (Musters and Bouten, 2000). Note that the two approaches have indeed some formal links with each other (Couvreur et al., 2012; Javaux et al., 2008).

The complexity of RWU modelling is highly related to the uneven root distribution in the vertical and radial directions (Gong et al., 2006). This variability is partly induced by heterogeneities in the soil and localized soil compaction caused by both cultivation and irrigation patterns (Jones and Tardieu, 1998) that in turn cause heterogeneous water and nutrient distribution. Consequently, there is a clear need for the development of novel RWU modelling approaches (Couvreur et al., 2012; Feddes et al., 2001; Raats, 2007; Jarvis, 2011;), as well as for accurate measurements techniques of soil water content and RWU dynamics.

In particular, soil moisture measurements are of paramount importance to calibrate RWU models. 85 Traditionally, and especially beneath irrigated crops, soil moisture has been determined using 86 87 methods such as neutron probes, TDR or capacitance systems. As these traditional techniques are point measurements, they do not provide sufficient information for reliable mass balance 88 assessments; therefore our understanding of RWU as a spatially distributed system remains 89 90 fundamentally limited. In this respect the understanding of soil as a spatially heterogeneous system shares fundamental limitations with most of earth sciences. Therefore much can be learnt looking 91 at similar research fields. 92

Geophysical methods have long been established for the imaging of the soil subsurface at a variety
of scales, from large scale mining exploration (e.g. Parasnis, 1973) to the very small scale of soil

95 mapping (e.g. Allred et al., 2008). The past twenty years, in particular, have seen the fast development of techniques that are useful in identifying structure and dynamics of the near surface, 96 with particular reference to hydrological applications. This realm of research goes under the 97 general name of hydro-geophysics (Binley et al., 201; Rubin and Hubbard, 2005, Vereecken et al., 98 2006) and covers a wide range of applications from flow and transport in aquifers (e.g. Kemna et 99 100 al., 2002; Perri et al., 2012) to the vadose zone (e.g. Daily et al., 1992), from catchment (e.g. Weill et al., 2013) and hillslope characterization (Cassiani et al., 2009a) to agriculture and eco-101 hydrological processes (Boaga et al., 2014; Ursino et al., 2014). 102

Possibly the most interesting results have been obtained when hydro-geophysical data have been coupled with distributed hydrological model predictions. The degree of integration of data and model range from trial and error calibration (e.g. Binley et al., 2002) to full data assimilation (e.g. Hinnell et al., 2010), but in all cases the availability of spatially extensive (and time intensive) data greatly improve the models' capability to identify within narrow ranges the relevant governing parameters, that in turn are of practical interest for hydrological predictions.

Relatively few hydro-geophysical applications, though, have been focussed on plant root system 109 110 characterization (e.g. al Hagrey et al, 2007; al Hagrey and Petersen, 2011; Javaux et al., 2008; 111 Jayawickreme et al., 2008, Werban et al., 2008), often limiting the analysis to a tentative 112 identification of the main root location and extent. Electrical soil properties are a clear indication of soil moisture content distribution and electrical and electromagnetic methods have been used to 113 114 identify the effect of root activity (e.g. Cassiani et al., 2012; Shanahan et al., 2015). In particular, 115 ERT has been used to characterize root water uptake and root system (Garré et al., 2011; Michot et al., 2001; Michot et al., 2003; Srayeddin and Doussan, 2009). Amato et al., (2009; 2010) tested the 116 ability of 3D-ERT for quantifying root biomass on herbaceous plants. Beff et. al., (2013) used 3D-117

118 ERT for monitoring soil water content in a maize field during late growing seasons. Boaga et al.,119 (2013) and Cassiani et al. (2015) demonstrated the reliability of the method in apple orchards.

In this paper we aim at applying hydro-geophysical techniques, with a combination of 120 121 measurements and modelling, to a tree root system. This approach has, to the best of our knowledge, not been presented and analysed yet. In particular, we present the application of the 122 time-lapse non-invasive 3D electrical resistivity tomography (ERT) to monitor soil-plant 123 interactions in the root zone of an orange tree located in the Mediterranean semi-arid Sicilian 124 (South Italy) context. The subsoil dynamics, particularly influenced by irrigation and RWU, have 125 126 been characterized by the 3D ERT measurements coupled with plant transpiration through sap flow measurements. The information contained in the ERT measurements in terms of vadose zone water 127 128 dynamics was exploited by comparing the field results against a 1-D vadose zone model.

129 The specific goals of this paper are

- (a) to study the feasibility of a small scale monitoring of root zone processes using time-lapse
 3D ERT;
- (b) to assess the value of the data above for a quantitative description of hyrological processesat the tens of centimeter scale;
- (c) to interpret these data with the aid of a physical hydrological model, in order to derive alsoinformation on the root zone physical structure and its dynamics.

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137 2. SITE DESCRIPTION

The experiment was conducted in a 20-hectar orange orchard, planted with about 20 year-old trees (*Citrus sinensis*, cv Tarocco Ippolito) (Figure 1). The field is located in Lentini (Eastern Sicily, Lat. 37°16'N, Long. 14°53'E) in a Mediterranean semi-arid environment, characterized by an annual average precipitation of around 550 mm, very dry summers and average air temperature of 7°C in 142 winter and 28°C in summer. The site presents conditions of crop homogeneity, flat slope, dominant wind speed direction for footprint analysis and quite large fetch that are ideal for 143 micrometeorological measurements. The planting layout is 4.0 m \times 5.5 m and the trees are drip 144 irrigated with 4 in-line drippers per plant, spaced about 1 m, with 16 L h⁻¹ of total discharge (4 L h⁻¹ 145 ¹ per dripper); the crop is well-watered by irrigation supplied every day from May to October, with 146 irrigation timing of 5 h d⁻¹. The study area has a mean leaf area index (LAI) of about 4 m² m⁻², 147 measured by a LAI-2000 digital analyser (LI-COR, Lincoln, Nebraska, USA). The LAI values are 148 spatially averaged and are referred to the ERT measurement period (October 2013). In the specific 149 150 case of a mature orange orchard, LAI values result fairly constant in time in the region of interest.

151 The mean PAR (photosynthetic active radiation) light interception was 80% within rows and 50% 152 between rows; the canopy height (h_c) is 3.7 m.

The soil characterization was performed via textural and hydraulic laboratory analyses, according 153 to the USDA standards. The area, covered by mature orange orchards, was divided into regular 154 grids, each having a 18×32 m² area, where undisturbed soil cores (0.05 m in height and 0.05 m in 155 diameter) were collected at the 0-0.05 m and 0.05-0.10 m depths for a total of 32 sampling points 156 and 64 soil samples. The undisturbed soil cores were used to determine the soil bulk density, ρ_b 157 (Mg m⁻³) and the initial water content, θ_i (m³m⁻³), i.e. the θ value at the time of the field campaign. 158 A total of 32 disturbed soil samples were also collected at the 0-0.05 m depth to determine the soil 159 160 textural characteristics using conventional methods following H₂O₂ pre-treatment to eliminate organic matter and clay deflocculation using sodium metaphosphate and mechanical agitation (Gee 161 162 and Bauder, 1986). Three textural fractions according to the USDA standards, i.e. clay (0-2 µm), silt (2-50 µm) and sand (50-2000 µm), were used in the study to characterize the soil (Gee and 163

Bauder, 1986). Most soil textures (i.e. 27 out of 32) were loamy sand and the remaining textureswere sandy loam.

An undisturbed soil sample was collected from the surface soil layer (0-0.05 m depth) at each sampling location (sample size, N = 32), using stainless steel cylinders with an inner volume of 10⁻¹ 4 m³ to determine the soil water retention curve. For each sample, the volumetric soil water content at 11 pressure heads, *h*, was determined by a sandbox (h = 0.01, 0.025, 0.1, 0.32, 0.63, 1.0 m) and a pressure plate apparatus (h = 3, 10, 30, 60, 150 m). For each sample, the parameters of the van Genuchten (1980, vG) model for the water retention curve with the Burdine (1953) condition were determined (Aiello et al., 2014).

Three soil water content profiles are measured in the field using water content reflectometers 173 174 (TDR, Time Domine Reflectometry), since 2009. Calibrated Campbell Scientific CS616 water content reflectometers (±2.5% of accuracy) were installed to monitor every 1 h the changes of 175 volumetric soil water content ($\Delta \theta$). The TDR probe installation was designed to measure soil water 176 177 content variations with time in the soil volume afferent to each plant. The TDR probes location is considered well suited with the specific characteristics of the micro-irrigation systems used in the 178 area and the textural soil main features. For each location the TDR equipment consists of two 179 sensors inserted vertically at 0.20 and 0.45 m depth and of two sensors inserted horizontally at 0.35 180 m depth, with 0.20 m in between. The water content reflectometer consists of two stainless steel 181 182 rods connected to a printed circuit board. When the probe rods were inserted vertically into the soil surface they gave an indication of the water content in the upper 20-25 cm of soil. The probes 183 installed horizontal to the surface were used to detect the passing of wetting fronts of water fluxes. 184

185 The data that are discussed here (see results section) correspond to the TDR probes located at 186 about 1.5 m from the orange tree we monitored with ERT. Hourly meteorological data (incoming short-wave solar radiation, air temperature, air humidity, wind speed and rainfall) are acquired by an automatic weather station located about 7 km from the orchard and managed by SIAS (Agro-meteorological Service of the Sicilian Region). For the dominant wind directions, the fetch is larger than 550 m. For the other sectors the minimum fetch is 400 m (SE).

192 **3. METHODOLOGY**

193 3.1 Micrometeorological measurements

The experimental site is equipped with Eddy Covariance (EC) systems mounted on a 194 micrometeorological fluxes tower (Figure 1). Continuous energy balance measurements have been 195 since 2009. In particular, net radiation (R_n , W m⁻²) is measured with two CNR 1 Kipp&Zonen 196 (Campbell Scientific Ltd) net radiometers at a height of 8 m. Soil heat flux density (G, W m⁻²) is 197 measured with three soil heat flux plates (HFP01, Campbell Scientific Ltd) placed horizontally 198 0.05 m below the soil surface. Three different measurements of G were selected: in the trunk row 199 200 (shaded area), at 1/3 of the distance to the adjacent row, and at 2/3 of the distance to the adjacent 201 row. The soil heat flux is measured as the mean output of three soil heat flux plates. Data from the 202 soil heat flux plates is corrected for heat storage in the soil above the plates.

The air temperature and the three wind speed components are measured at two heights, 4 and 8 m, using fine wire thermocouples (76 µm diameter) and sonic anemometers (Windmaster Pro, Gill Instruments Ltd, at 4m, and a CSAT, Campbell Sci., at 8 m). A gas analyzer (LI-7500, LI-COR) operating at 10 Hz was installed at 8 m. The raw data are recorded at a frequency of 10 Hz using two synchronized data loggers (CR3000, Campbell Sci.).

The Eddy Covariance measurement system and the data processing followed the guidelines of the standard EUROFLUX rules (Aubinet et al., 2000). A data quality check was applied during the post processing together with some routines to remove the common errors: running means for detrending, three angle coordinate rotations and de-spiking. Stationarity and surface energy closure
were also checked (Kaimal and Finningan, 1994).

Low frequency measurements are taken for air temperature and humidity (HMP45C, Vaisala), wind speed and direction (05103 RM Young), and atmospheric pressure (CS106, Campbell Scientific Ltd) at 4, 8 and 10 m.

The freely distributed TK2 package (Mauder and Foken, 2004) is used to determine the first and second order statistical moments and fluxes on a half-hourly basis following the protocol used as a comparison reference described in Mauder et al. (2007).

219 3.2 Sap flow measurements

Heat-pulse techniques can be used to measure sap flow in plant stems with minimal disruption to the sap stream (Cohen et al., 1981; Green and Clothier, 1988; Swanson and Whitfield, 1981). The measurements are reliable, use inexpensive technology, provide a good time resolution of sap flow, and they are well-suited to automatic data collection and storage. Sequential or simultaneous measurements on numerous trees are possible, permitting the estimation of transpiration from whole stands of trees.

Measurements of water consumption at tree level (T_{SF}) have been taken using the HPV (Heat Pulse Velocity) technique that is based on the measurement of temperature variations (ΔT), produced by a heat pulse of short duration (1-2 s), in two temperature probes installed asymmetrically on either side of a linear heater that is inserted into the trunk. For HPV measurements, two 4 cm sap flow probe with 4 thermocouples embedded (Tranzflo NZ Ltd., Palmerston North, NZ) were inserted in the trunks of the trees, belonging to the area of footprint of the micrometeorological eddy covariance tower. The probes were positioned at the North and South sides of the trunk at 50 cm from the ground and wired to a data-logger (CR1000, Campbell Sci., USA) for heat-pulse control and measurement; the sampling interval was 30 min. The temperature measurements are obtained by means of ultra-thin thermocouples that, once the probes are in place, are located at 5, 15, 25 and 45 mm within the trunk.

Data have been processed according to Green et al. (2003b) to integrate sap flow velocity over sapwood area and calculate transpiration. In particular, the volume of sap flow (Q_{stem}) in the tree stem is estimated by multiplying the sap flow velocity by the cross sectional area of the conducting tissue. To this purpose, fractions of wood (F_M =0.48) and water (F_L =0.33) in the sapwood were determined on the trees where sap flow probes were installed. Wound-effect correction (Consoli and Papa, 2013; Green et al., 2003b; Motisi et al., 2012) was done on a per-tree basis. Crop transpiration data are available at the study site since 2009.

244 3.3 Electrical resistivity tomography (ERT)

245 The key technique used to monitor the soil moisture content distribution in the volume surrounding the orange tree is electrical resistivity tomography (ERT – e.g. Binley and Kemna, 2005). In 246 247 particular, we installed a three-dimensional ERT system, consisting of 48 buried electrodes placed 248 on 4 instrumented micro-boreholes, with 12 electrodes each (see Figure 2). The electrodes are 249 made of stainless steel wound around a one-inch PVC pipe, and are spaced 10 cm along the pipe (see inset in Figure 2), thus the shallowest and the deepest are respectively at 0.1 m and 1.2 m 250 251 below the surface. Each electrode is made of a plate 3 cm wide. The boreholes are placed at the 252 vertices of a square, having a side of 1.3 m, that has the orange tree at its centre, and were inserted by percussion with the help of a pre-drilling with a smaller diameter in order to avoiding the 253 254 disturbance of the electrical flow. The electrical contact is excellent for all 48 buried electrodes, as checked before each measurement. The 4 boreholes are water tight and in tight contact with the 255

256 soil, so they cannot act as pathways for preferential water infiltration. We focused our attention to an area slightly smaller than the square defined by the boreholes, in order to avoid the inevitable 257 disturbance caused by borehole installation (slightly compacting the surrounding soil). The system 258 259 is completed by 24 electrodes at the ground surface, placed along a square grid of about 0.21 m side, covering the 1.3 m x 1.3 m square at the surface (Figure 3): this setup allows a homogeneous 260 261 coverage of the surface of the control volume. The chosen acquisition scheme was a skip-zero 262 dipole-dipole configuration, i.e. a configuration where the current dipoles and potential dipoles are both of minimal size, i.e. they consist of neighbouring electrodes e.g. along the boreholes. This 263 264 setup ensures maximal spatial resolution (as good as the electrode spacing, at least close to 265 electrodes themselves) provided that the signal/noise ratio is sufficiently high. The data quality is assessed using a full acquisition of reciprocals to estimate the data error level (see e.g., Binley et 266 al., 1995; Monego et al., 2010). Consistently, we used for the 3D data inversion an Occam 267 approach as implemented in the R3 software package (Binley, 2014) accounting for the error level 268 269 estimated from the data themselves. The relevant three-dimensional computational mesh is shown in Figure 3. At each time step, about 90-95 % of the dipoles survived the 10% reciprocal error 270 271 threshold. In order to build a time-consistent data set, only the dipoles surviving this error analysis 272 for all time steps were subsequently used, reducing the number to slightly over 90% of the total. 273 The absolute inversions were run using the same 10% error level. Time-lapse inversions were run at a lower error level equal to 2 % (consistently with the literature – e.g., Cassiani et al., 2006). 274 275 We conducted repeated ERT measurements using the above apparatus for about two days, starting

276 on October 2, 2013 at 11:00 am, and ending the next day at about 16:00. The schedule of the 277 acquisitions and the irrigation times is reported in Table 1. Note that the background ERT survey 278 was acquired on October 2 at 11:00 before the first irrigation period was started, so that all changes 279 caused by irrigation and subsequent evapotranspiration can be referred to that instant. Note that prior to October 2, 2013, irrigation had been suspended for at least 15 days. Note also that only one dripper – with a flow of about 4 l/h – is located at the surface of the control volume defined by the ERT setup (Figure 3).

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4 4. RESULTS AND DISCUSSION

The paper presents results derived from both short term (2 days) and long term monitoring. The micrometeorological data set (including the measurements of the energy balance components) and the sap flow data are available since 2009. ERT measurements were carried out only during a 2day period, but the state of the system at the time of the ERT measurements clearly depends on the past forcing acting on the system. In order to fully exploit the information content of this dataset, we aimed at comparing data against simulations, as much as possible in a quantitative manner.

291 The ERT monitoring as described in Table 1 produced two clear results:

- (1) The initial conditions (11:00 a.m. of October 2, before irrigation starts) around the tree 292 293 show a very clear difference in electrical resistivity in the top 40 cm of soil with respect to 294 the rest of the volume (Figure 4). Specifically, the resistivity of the top layer ranges around 40-50 Ohm m, while the lower part of the profile is about one order of magnitude more 295 296 conductive (about 5 Ohm m). As no apparent lithological difference is present at 40 cm 297 depth (see also laboratory results below) we attributed this difference to a marked difference in soil moisture content. This was confirmed by all following evidence (see 298 299 below).
- 300 (2) The resistivity changes as a function of time, during the two irrigation periods, during the
 301 night interval, and afterwards, all show essentially the same pattern, with relatively small
 302 (but still clearly measureable) changes (Figure 5). Two zone are identifiable: (a) a shallow

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zone (top 10-20 cm) where resistivity decreases with respect to the initial condition; and (b) a deeper zone (20-40 cm) where resistivity increases.

Qualitatively, both pieces of evidence can be easily explained in terms of water dynamics governed 305 306 by precipitation, irrigation and root water uptake. Specifically, the shallower high resistivity zone in Figure 4 can be correlated to a dry region where root water uptake manages to keep soil moisture 307 308 content to minimal values, as an effect of the entire summer strong transpiration drive. The dynamics in Figure 5, albeit small compared to the initial root uptake signal in Figure 4, still 309 confirm that the top 40 cm is house to a strong root activity, to the point that irrigation cannot raise 310 311 electrical conductivity of the shallow zone (10-20 cm) by no more than some 20%, and the roots 312 manage to make the soil even drier (with a resistivity increase by some 10%) in the 20-40 cm depth layer (Figure 5). Note that, in general, resistivity changes of the type here observed cannot be 313 314 uniquely associated to soil moisture content changes, as pore water conductivity may play a key role (e.g. Boaga et al., 2013; Ursino et al., 2014). However, in the particular case at hand, care was 315 316 taken to analyze the electrical conductivity of both the water used for irrigation and the pore water, 317 purposely extracted at about 50 cm depth. Both waters showed an electrical conductivity value in the range of 1300 µS/cm (thus fairly high, fact that explains the overall small soil resistivity 318 319 observed at the site). Therefore in this particular case we can exclude pore water conductivity 320 effects in the observed dynamics of the system. Once again it must be stressed that this is rather the 321 exception than the rule.

A laboratory-based method was adopted for obtaining "unaltered" soil pore water through a column displacement technique (Knight et al., 1998). In particular, Rhizon soil moisture samplers (Cabrera, 1998) were used; they represent one of the latest developments in terms of tension samplers, where it is necessary to apply a suction to withdraw pore water with a vacuum tube (Tye et al., 2003).

The qualitative evidence above is, however, not very surprising and not particularly informative: the root activity dries the soil, this is not a discovery. Things become more interesting if we can translate the ERT data into quantitative estimates of soil moisture content, and if we can use these data to calibrate hydrological models of the root zone.

To this end, we tested Bulgherano soil samples in the laboratory to obtain a suitable constitutive 331 332 relationship linking moisture content and resistivity, given the know pore water conductivity that was reproduced for the water used in the laboratory. All measurements were conducted using 333 cylindrical Plexiglas cells equipped with a four-electrode configuration designed to allow for 334 sample saturation and de-saturation with no sample disturbance, using an air injection apparatus at 335 336 one end and a ceramic plate at the opposite end. samples. The air entry pressure of the ceramic is 1 bar, thus during all the experiments the plate remained under full water saturation, 337 338 while allowing water outflow during de-saturation. At each de-saturation step, the electrical conductivity of the sample was measured under temperature controlled conditions using a ZEL-339 340 SIP04 impedance meter (Zimmermann et al. 2008). A completed description of the setup is 341 given by Cassiani et al. (2009b).

342 Figure 6 shows two example experimental results on samples from two different depths. Note how 343 in a wide range of soil moisture content (roughly from 5% to saturation) the two curves in Figure 6 lie practically on top of each other. The same applies for all tested samples. Note also that, even 344 though some samples show the effect of the conductivity of the solid phase (through its clay 345 346 fraction) at small saturation (see sample from 0.4 m in Figure 6) still the effect is small as it appears only at soil moisture smaller than 3-4%. Therefore we deemed unnecessary to resort to 347 constitutive laws that represent this solid phase effect, such as Waxman and Smits (1968) that has 348 349 been used for similar purposes elsewhere (e.g. Cassiani et al., 2012) and we adopted a simpler 350 Archie's (1942) formulation. Consequently we translated resistivity into moisture content using the

following relationship calibrated on the laboratory data, using a water having the above mentionedelectrical conductivity:

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$$\theta = \frac{4.703}{\rho^{1.12}}$$
(1)

where θ is volumetric soil moisture content (dimensionless) and ρ is electrical resistivity (in Ohm 354 m). The relationship (1) allows a direct translation of the 3D resistivity distribution to a 355 corresponding distribution of volumetric soil moisture content. However, it has long been 356 357 established that inverted geophysical data may bring with them enough distortion of the true physical parameter field (Day-Lewis et al., 2005) as to induce violations of elementary physical 358 principles, such as mass balance during tracer test monitoring experiments (e.g. Singha and 359 360 Gorelick, 2005). This may cause substantial problems, particular when the use of data is expected to shift from a qualitative interpretation to a quantitative use in terms of data assimilation into 361 hydrological models. For this reason, coupled versus uncoupled approaches have been proposed 362 and discussed (Hinnell et al., 2010) even though their superiority seems to depend on the specific 363 problem, as the information content of data even in a tradition, inverted approach may be sufficient 364 365 (Camporese et al., 2011, 2014). Indeed, the geometry we are considering here is very effective to 366 reconstruct the mass balance of irrigated water, as this comes as a quasi-one dimensional infiltration from the top, where in addition electrodes are located. The geometry is similar to 367 368 the one used, e.g., Koestel et al. (2008) where mass balance was verified by comparison against very detailed TDR data collected in a lysimeter. In spite of these considerations, we decided to still 369 limit ourselves to analyzing the data variation principally as a function of depth, lumping the data 370 371 horizontally by averaging estimated moisture content along two-dimensional horizontal planes. Note that the dataset may lend itself to more complex analyses such as the one proposed by Manoli 372 et al. (2014), especially if used in the context of a formal Data Assimilation, but we felt that one 373

374 such an endeavor would exceed the scope of the current paper and deserves an ad-hoc space. Note 375 also that the ERT field evidence both in terms of background (Figure 4) and time-lapse evolution 376 (Figure 5) of moisture content confirm the hypothesis that, within the control volume, the 377 distribution of water in the soil is largely one-dimensional as a function of depth.

The data, once condensed in this manner, lend themselves more easily to a comparison with the 378 379 results of infiltration modeling. We implemented a one-dimensional finite element model based on a Richards' equation solver (Femwater – details of this classical model are given by Lin et al., 380 1997), simulating the central square meter of the ERT monitored control volume, down to a total 381 382 depth of 2 meters (much below the depth of the ERT boreholes), where we assumed that the water table is located (Dirichlet boundary condition). We applied at the top of the soil column a 383 Neumann boundary condition consistent with the flux coming from irrigation that pertains the 384 385 control volume (basically, the water coming from a single dripper). As Femwater is a 3D simulator, the soil column is also bounded laterally by no-flow conditions, with the exception of the top 40 386 387 cm where we applied laterally a Neumann condition simulating the root water uptake (see below 388 for details).

We considered only the central part of the ERT-controlled volume (1 m x 1 m) thus excluding the regions too close to the boreholes that, even though benefitting from the best ERT sensitivity, might have been altered from a hydraulic viewpoint by the drilling and installing operations. Correspondingly we averaged horizontally the ERT data only in this central region.

A very fine vertical discretization (0.01 m) and time stepping (0.01 h) ensures solution stability. The porous medium is homogeneous along the column and parameterized according to the Van Genuchten (1980) model. The relevant parameters had been derived independently from laboratory and field measurements, the latter particularly relevant for the definition of a reliable in situ saturated hydraulic conductivity estimate. The parameters used for the simulations are: residual

moisture content $\theta_r = 0$, porosity $\theta_s = 0.54$, $\alpha = 0.12$ 1/m, n = 1.6, saturated hydraulic conductivity 398 $K_s = 0.002$ m/h. We acknowledge that a more complete sensitivity analysis concerning the impact 399 of the individual parameters would be beneficial, but this should be performed in a complete Monte 400 401 Carlo manner in order to exclude identification trade-offs between the Van Genuchten parameters, 402 the depth of the water table (known with some uncertainty) and the fluxes from irrigation, 403 precipitation and evapotranspiration. However we feel that this endeavour shall be conducted also with regard to the effective 3D spatial distribution of active roots, and is currently the subject of 404 405 ongoing research.

The remaining elements of the predictive modelling exercise are initial and boundary conditions. 406 407 As we focused primarily our attention on reproducing the state of the system at background conditions, we set the start of the simulation at the beginning of the year (1/1/2013), and we 408 409 assumed for that time a condition drained to equilibrium. Given the van Genuchten parameters we 410 used and the depth of the water table, this corresponds to a fairly wet initial condition. We verified 411 a posteriori that moving the initial time back of one or more years did not alter the predicted results at the date of interest (October 3, 2013). The dynamics during the year are sufficient to bring the 412 413 system to the real, much drier condition in October. The forcing conditions on the system are all known: (a) irrigation is recorded, and only one dripper pertains to the considered square meter; (b) 414 precipitation is measured; (c) sap flow is measured. Direct evaporation from the square meter of 415 soil around the stem is neglected, considering the dense canopy cover and the consequent limited 416 417 radiation received. Only one degree of freedom is left to be calibrated, i.e. the volume from which the roots uptake water. Thickness of the active root zone was estimated from the time-lapse 418 419 observations (Figure 5), and fixed to the top 0.4 cm after checking that limiting the root uptake to the 0.2 m to 0.4 m zone would produce results inconsistent with observations in the top 0.2 m. 420 421 Therefore only the surface area of the root uptake zone remains to be estimated. We used the 422 predictive model as a tool to identify the extent of this zone, that is of critical interest also for423 irrigation purposes.

Figure 7 shows the results of the calibration exercise. It is apparent that the total areal extent of the root uptake zone has a dramatic impact on the predicted moisture content profiles, as it scales the amount of water subtracted from the monitored square meter considered in the calibration. Even relatively small changes (+/-15%) of the root uptake area produce very different soil moisture profiles. The value that allows a good match of the observed profile is 1.75 m^2 , while for areas equal to 1.5 m^2 and 2 m^2 the match is already unsatisfactory, leading respectively to underestimation and overestimation of the moisture content in the profile.

Another important fact that is apparent from Figure 7 is that the estimated soil moisture in the 431 shallow zone (roughly down to 0.4 m) is very small as an effect of root water uptake. However this 432 dry zone must have a limited areal extent $(1.75 \text{ m}^2, \text{ corresponding to a radius of about } 0.75 \text{ m from})$ 433 the stem of the tree). Indeed this is indirectly confirmed by the soil moisture evolution measured by 434 435 TDR. Figure 8 shows the TDR data from three probes located about 1.5 m from the monitored tree (thus outside our estimated root uptake zone). The signal coming from the irrigation experiment of 436 437 October 2, 2013 is very apparent with an increase in moisture content of all three probes, located at 438 different depths. Note that before this experiment the system had been left without irrigation for 439 about two weeks. The corresponding effect on the TDR data is apparent: all three probes show a decline of moisture content during the day, with pauses overnight. The decline is more pronounced 440 441 in the 0.35 m TDR probe, that lies at a depth we estimated to be nearly at the bottom of the RWU 442 zone, and less pronounced above (0.2 m) and below (0.45 m). Note also that the TDR probes are close to another dripper, lying outside of the ERT controlled volume (the drippers are spaced 1 m 443 along the orange trees line, with the trees about 4 m from each other) thus they reflect directly the 444 infiltration from that dripper. However, at all three depths the moisture content is much higher than 445

446 measured in the ERT-controlled block closer to the tree. This can be explained with the fact that in that region the root uptake is minimal or totally absent, while the decline of moisture content in 447 time may well be an effect of water being drawn to the root zone by lateral movement induced by 448 449 the very strong capillary forces exerted by the dry fine grained soil in the active root zone closer to the tree. In order to clarify the impact of these results on our understanding of the system, we show 450 451 the location of the trees, of the TDR probes and of the drippers in Figure 9, where we also sketch 452 the best estimate for the areal extent of the RWU zone. This Figure clearly highlights how critical the information provided by ERT actually is. The scale at which RWU takes place is smaller 453 454 (meter scale) than expected and often assumed when it comes to designing and implementing a 455 field monitoring system. This has dramatic consequences in terms of how reliable conclusions can be drawn if such a small scale processes are neglected. Consider, e.g., what type of conclusions 456 457 could be drawn on the basis of TDR data alone (Figure 8) in light of the field situation as depicted in Figure 9. The single, most important message that shall be conveyed by this paper is a warning 458 459 to be particularly attentive to small scale processes in soil-plant-atmosphere interactions, even in 460 regular agricultural landscapes.

461

462 **5. CONCLUSIONS**

Near surface geophysics is strongly affected by both static and dynamic soil/subsoil characteristics. This fact, if properly recognized, is potentially full of information on the soil/subsoil structure and behaviour. The information is maximized if geophysical data are collected in time-lapse mode. In the case of interactions with vegetation, its role should be properly modelled, and such models can be constrained by means (also) of geophysical data. This case study demonstrates that 3D ERT is capable of characterizing the pathways of water distribution, and provides spatial information on root zone suction regions. The integration of modelling and data has proven, once again, a key 470 component of this type of hydro-geophysical studies, allowing us to draw quantitative results of 471 practical interest. In this case we had available a wealth of quantitative information about transpiration and soil moisture content that allowed the definition of the volume of soil affected 472 473 by the RWU activity. This has obvious consequences for the possible improvement of irrigation strategies, as it is apparent how the monitored orange tree essentially drives water from 1 to 2 474 475 drippers out of the 4 total that should pertain to its area in the plantation. This means that it is very 476 likely that half of the irrigated water is indeed lost to deeper layers and brings no contribution to the plants. More advanced uses of this type of data are now considered, especially linking soil 477 478 moisture distribution with plant physiological response and active root distribution in the soil. In 479 the long run studies of this type may give a fundamental contribution to our understanding of soilplant-atmosphere interactions also in view of facing challenges coming from climatic changes. 480

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Acquisition #	Starting time	Ending time	Irrigation schedule	Date
0 (background)	10:40	11:00		
1	12:00	12:20		
2	13:00	13:20	11:30 to 16:30	
3	14:15	14:35	4 l/h from	October 2, 2013
4	15:00	15:20	each dripper	
5	16:00	16:20		
6	17:00	17:20		
7	10:15	10:35		
8	11:05	11:25		
9	12:00	12:20	7:00 to 12:00	
10	13:00	13:20	4 l/h from	October 3, 2013
11	14:00	14:20	each dripper	
12	15:00	15:20		
13	15:45	16:05		

 Table 1: times of acquisitions and irrigation schedule



- **Figure 1:** Bulgherano experimental site: the Eddy Covariance (EC) tower and a Heat Pulse (HP)
- 712 Sap Flow installation on an orange tree.



Figure 2: 3D ERT apparatus installed around one orange tree. The system is composed of four
micro-boreholes carrying 12 electrodes each (see inset) and 24 surface electrodes – see text and
Figure 3 for geometry details.



Figure 3: Electrode geometry around the orange tree and 3D mesh used for ERT inversion.



Figure 4: cross-sections of the ERT cube corresponding to the background acquisition of October
2, 2013, 11:00 a.m. Note the very strong difference in electrical resistivity between the top 40 cm
(above 50 Ohm) and the rest of the domain. The resistivity distribution is essentially onedimensional with depth, with very limited horizontal variations.



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Figure 5: (A) time series of sap flow (black line) and EC-derived total evapotranspiration (blue lines), both normalized in mm assuming an area of 20 m² pertaining to the orange tree monitored with ERT. Time is given in hours from midnight of October 2. The two irrigation periods are shown by the blue bars. (B) 3D ERT images of resistivity change with respect to background at two selected time instants shown by the arrows in (A); the volumes corresponding to increase and decrease of resistivity above and below certain thresholds (80% and 110%) are shown in separate panels, for clarity.



Figure 6: experimental relationships between resistivity and moisture content determined in the lab
on samples taken at two different depths at the Bulgherano site, using water having the same
electrical conductivity measured in the pore water in situ.



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Figure 7: results of 1D Richards' equation simulations of the entire year 2013 till October 3, 11:00 a.m., i.e. in correspondence of the background ERT acquisition (the thick black line represents the resulting estimated moisture content profile obtained from averaging horizontally the central square meter of the ERT control volume). The different simulated curves correspond to different assumed areas of root water uptake, and show how 1.75 m2 is the area that allows to match the observed real profile with good accuracy. Note also the high sensitivity of the results to the estimated root uptake area.



Figure 8: moisture content time series from three TDR probes located about 1.5 m from the ERTmonitored tree. The signal coming from the irrigation experiment of October 2, 2013 is very clear.

763 Before this experiment the system had been left without irrigation for about two weeks.



Figure 9: scheme of the experimental field with the location of the main sensors. The radius of the

root water uptake zone, assumed to be circular, is equal to about 0.75 m.

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