# A mechanistic model for electrical conduction in soil-root continuum: a virtual rhizotron study

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- 19 Abstract

20 Electrical Resistivity Tomography (ERT) has become an important tool to study soil water fluxes

in cropped field. ERT results translate to water content via empirical pedophysical relations that 21 take soil physical properties into account, usually ignoring the impact of roots. Studies show that 22 high root dense soils behave quite differently than less root dense soils in terms of bulk electrical 23 conductivity. Yet, we do not completely understand the impact of root segments on the ERT 24 measurements. In this numerical study, we coupled an electrical model with a plant-soil water flow 25 model to investigate the impact of plant root growth and water uptake on the ERT virtual 26 experiment. The electrical properties of roots were explicitly accounted for in the finite element 27 mesh and we obtained the electrical conductivities of root segments by conducting specific 28 experiments on real maize plants. The contrast between electrical conductivity of roots and soil 29 depends on factors such as root density, irrigation, root age, and root water uptake pattern. Root 30 31 growth and water uptake processes thus affect this contrast together with the soil electrical properties. Model results indicate a non-negligible anisotropy in bulk electrical conductivity 32 induced by root processes. We see a greater anisotropy in a sandy medium when compared to a 33 34 loamy medium. We find that the water uptake process dominates the bulk electrical properties. The 35 Gauss-Newton type ERT inversion of virtual rhizotron data demonstrate that, when root-soil electrical conductivity contrasts are high, it can lead to error in water content estimates since the 36 electrical conductivity is partly due to root. Thus, incorporating the impact of root in the

pedophysical relations is very important to interpret ERT results directly as water content.

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## 43 **1 Introduction**

44 Understanding root water uptake and associated nutrients is critical for crop management (e.g. Gregory et al. 2005) but remains a challenging task due to the inherent difficulty to collect 45 46 observations in the soil (e.g. de Dorlodot et al., 2007). Geophysical monitoring of soil-root system water fluxes have received growing interest in the past decades to tackle this challenge. In 47 particular, in this paper, we will investigate the potential of Electrical Resistivity Tomography 48 (ERT) (Michot et al., 2003; Paglis, 2013). This method aims at retrieving the 2D or 3D distribution 49 50 of the electrical conductivity ( $\sigma$ ) or its inverse resistivity in the soil. The electrical conductivity is then related to the variable of interest (for instance the soil water content SWC) through a 51 pedophysical or petrophysical relationship. 52

53 In cropped fields, ERT has been increasingly used for monitoring soil water content (SWC) (Beff et al., 2013; Brillante et al., 2016; De Carlo et al., 2015; Garrè et al., 2011; Michot et al., 2003; 54 Srayeddin and Doussan, 2009; Vanella et al., 2018). More recently, ERT-estimated water content 55 was used for phenotyping root systems at field scale (Whalley et al., 2017). The authors monitored 56 changes in  $\sigma$  of the soil root zone in drying condition at different soil depths, which acted as a 57 proxy of root activity. However, apparent or bulk conductivity of a vegetated soil (potentially 58 containing roots), denoted by  $\sigma_{bulk}$ , is not only dependent on SWC but also on roots and their 59 impact on soil structure. Field experiments further show that the rooted zone soil behaves quite 60 61 different in terms of pedophysical relation as compared to soil containing no roots (Michot et al., 2016; Werban et al., 2008). Therefore ERT-monitored SWC in agricultural fields can be inaccurate 62 or misleading if we ignore the impact of root-related processes on the bulk conductivity of the soil-63 root continuum. 64

In the literature, various studies mention or even target the impact of roots on  $\sigma_{bulk}$ . In Fig. 1, we report values of bulk soil electrical conductivity without roots, denoted by  $\sigma_{bulk-soil}$ , and root segment electrical conductivity, denoted by  $\sigma_{root}$ . The ratio between  $\sigma_{root}$  and  $\sigma_{bulk-soil}$  is generally a function of plant species, soil type, SWC and solute concentration.

For a given species,  $\sigma_{root}$  is a function of root anatomy, which can be related to root age, root order or root diameter. In their study, Anderson and Higinbotham (1976), found that older maize root segments are electrically more conductive than younger roots. Their study was performed on excised root segments. They showed that the outer layer of the root segment (cortex) has very low 73 electrical resistance (~50 k $\Omega$ ) in the radial direction when compared to the axial direction (~600  $k\Omega$ ). By treating cortex and stele as concentric parallel conductors, the reported resistances, when 74 75 converted into conductivity are of the order  $\sigma_{root} \sim 0.05$  S/m. However, the electrical behavior of intact root segments embedded in the soil might be different as compared to excised segments. 76 77 Another study by Cao et al. (2010) reported that the root electrical resistance could be related to root properties such as surface area, number of lateral roots and root length. Studies on poplar roots 78 79 show that  $\sigma_{bulk}$  of the soil-root medium may increase or decrease with the increase in root mass density depending on the age of the plant (Al Hagrey, 2007; Zenone et al., 2008). On the other 80 hand,  $\sigma_{bulk-soil}$  depends on several factors, the most important being the porosity of the soil, the 81 electrical conductivity of the soil fluid ( $\sigma_w$ ), and SWC. In addition, loamy and clayey soils have a 82 surface conductivity that depends on mineral composition, SWC and  $\sigma_w$  (Friedman, 2005). 83

Literature on root electrical properties (Anderson and Higinbotham, 1976; Cao et al., 2010, 2011; 84 Ginsburg and Laties, 1973; Paglis, 2013) and pedophysical models for soils (Al Hagrey, 2007; 85 Amente et al., 2000; Bhatt and Jain, 2014; Friedman, 2005; Garrè et al., 2011; Laloy et al., 2011; 86 Werban et al., 2008; Wunderlich et al., 2013) suggest that if the contrast between  $\sigma_{root}$  and 87  $\sigma_{bulk-soil}$  is large enough, roots could have a measurable impact on ERT inversion results. In 88 89 addition there are studies that found a correlation between root length/mass density and electrical resistivity obtained from ERT (Amato et al., 2009; Rossi et al., 2011). These studies used 90 91 destructive methods to determine root length density and root biomass. However, to our knowledge, there are no detailed modeling efforts to study the effects of roots on electrical 92 93 conductivity of the bulk medium when monitoring SWC in cropped fields using the ERT method.

Beyond the impact of the electrical conductivity of root tissues, root-related processes like water 94 95 uptake, exudation or solute uptake will also affect the electrical properties of the rhizosphere, i.e. the soil zone in close proximity to root segments, thereby affecting the  $\sigma_{root}$  -  $\sigma_{bulk-soil}$  contrast. 96 The evolution of plant transpiration and root growth will also constantly impact the  $\sigma_{root}$  -97 98  $\sigma_{bulk-soil}$  patterns. Recent ERT experiments on orange orchard fields suggest that ERT results are more sensitive to root water uptake pattern (Vanella et al., 2018) than the presence of resistive 99 100 lignified roots. While this may be true for orange trees, we need a thorough study to investigate the 101 sensitivity of ERT results on the presence of different types of root that are more electrically conductive than soil. Therefore, to investigate the impacts of roots on ERT derived SWC, we 102

should take into account the root water uptake, soil heterogeneity, root specific electrical propertyalong with root growth.

105 To validate and quantify the impact of roots on ERT-derived SWC, we propose to simulate ERT on a virtual soil-root system. Al Hagrey and Petersen (2011) studied the impact of roots on ERT 106 107 imaging by using a root growth model (Wilderotter, 2003), however they ignored the inherent heterogeneity of  $\sigma_{root}$  and  $\sigma_{bulk-soil}$ . To understand the effect of root system connectivity and their 108 impact on SWC on  $\sigma_{bulk}$ , a model where roots are explicitly represented is needed. Explicit root 109 representation using an unstructured finite element mesh has been studied for water and nutrient 110 111 uptake processes (Tournier et al., 2015; Wilderotter, 2003), but to the best of our knowledge, no 112 such work exists for ERT simulations coupled to a plant-soil water flow model.

113 The objective of this study is to investigate how a transpiring growing plant might affect the ERT 114 estimate of SWC. We hypothesize that the  $\sigma$  contrast between the plant root system and the soil surrounding the roots (impacted by root, soil properties, and plant hydraulic boundary conditions) 115 116 together with the amount of roots will affect the ERT measurements and therefore ERT-derived quantities. In our work, we model the electrical conductivity of the soil-root system in a rhizotron 117 geometry with a fine spatial resolution for the roots using an unstructured mesh for the ERT 118 simulation. The root model includes transient transpiration, root growth and root and soil water 119 120 redistribution. We choose the maize root system for our study and exclude root exudation and solute uptake processes. We also study anisotropy in the electrical conductivity induced by root 121 122 growth and the water depletion pattern. An accurate electrical conductivity model of the soil-root system will improve our understanding of the electrical behavior of the soil-root zone and hence 123 will help us in improving the ERT method as a feasible and faster tool to monitor soil moisture in 124 vegetated land. This study is therefore a first step towards a thorough understanding of the impact 125 126 of roots on SWC monitoring using the ERT method.

# 127 2 Materials and Methods

Our numerical experiment consists of running a combination of highly detailed simulations representing the soil water fluxes in a planted 2-D rhizotron along with an ERT simulation. Root and soil electrical and hydraulic properties were explicitly accounted for and spatially distributed with a high resolution to study how root architecture and water uptake influence the ERT imaging results and the interpretation in terms of SWC. Fig. 2 summarizes the various steps described belowin a flow diagram.

#### 134 **2.1 Rhizotron/plant water flow model**

135 A two-dimensional root architecture was extracted from light transmission experiments on a real 136 rhizotron with a 21 days old maize species using the root image analyzing tool SMARTROOT (Lobet et al., 2011). The digitized root (see Fig.3) was then used for root water uptake modeling 137 using R-SWMS (Javaux et al., 2008). Since the root growth was monitored every day, ages were 138 easily assigned to each root segment. Root growth was simulated by updating the root system 139 architecture at each time step between the beginning (day 5) and the end (day 22) of the simulation. 140 141 Cyclic transpiration demand was imposed as top boundary condition for the root system. The daily 142 transpiration was supposed to linearly increase between the root emergence and the end of the study. At day 22, daily transpiration reached 25 cm<sup>3</sup>. 143

The root system is entirely contained in soil box whose length, thickness and depth were 22 cm, 1 cm, and 40 cm respectively (the corresponding reference axes are -11 < x < 11cm, -0.5 < y < 0.5 cm, -40 < z < 0 cm). In the scenario analysis, we considered both sandy and loamy soil types whose hydraulic properties were supposed to be perfectly represented by Mualem-van Genuchten equations (van Genuchten, 1980). Hydraulic parameters for both soils are given in Table 1.

The initial soil condition was a hydrostatic equilibrium with a saturated soil at the bottom of the rhizotron and root transpiration was the only source/sink term that allowed the total water content to change. R-SWMS (Javaux et al., 2008) uses the finite element method on a regular uniform grid to solve Richards equation in order to simulate three-dimensional water flow in the soil:

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$$\frac{\partial\theta}{\partial t} = \frac{\partial}{\partial x} \left( K \frac{\partial h}{\partial x} \right) + \frac{\partial}{\partial y} \left( K \frac{\partial h}{\partial y} \right) + \frac{\partial}{\partial z} \left( K \frac{\partial (h+z)}{\partial z} \right) - Sink$$
[1]

154 ,where  $\theta$  is the volumetric SWC, *h* the matrix head, *K* the isotropic hydraulic conductivity, *Sink* 155 is a sink term for root water uptake [cm<sup>3</sup> cm<sup>-3</sup> day<sup>-1</sup>], and *x*,*y* and *z* are the spatial coordinates. 156 Experimentally measured maize root hydraulic conductivities were used in the R-SWMS model, 157 in which they are age and type dependent (Couvreur et al., 2012; Doussan et al., 2006). Two-158 dimensional distributions of roots and of SWC were subsequently transformed into electrical 159 conductivity maps through appropriate bio-pedo electrical relations.

## 160 2.2 Electrical properties of plant root tissues and soils

To get insight into maize root electrical properties, we designed specific experiments on intact root 161 segments (Ehosioke et al., in preparation). First, we identified and separated the primary and brace 162 roots from maize plants grown in laboratory and were thoroughly washed with demineralized water 163 164 and dried with absorbent tissue. The electrical resistance of root segments was measured using a digital multimeter and were converted into electrical conductivity ( $\sigma_{root}$ ) by approximating the 165 root segment as a cylindrical geometry similar to Cao et al. (2010). The measurement direction of 166 root segments in Cao et al. (2010) is from root apex towards root collar while it is opposite in the 167 case of our experiment. We studied intact root segments as compared to excised root segments in 168 169 the studies of Cao et al. (2010) or Anderson and Higinbotham (1976) and investigated both primary and brace roots in the experiments. We examined variations of  $\sigma_{root}$  with respect to the segment 170 distance from the root collar and root cross-sectional area with a segment length of 4 cm. 171 172 Conductivity gel (Rodisonic, from Pannoc Nv/SA Belgium) was used to improve the electrical 173 contact between root segments and measuring electrodes. However, in the simulation model, only the variations of  $\sigma_{root}$  as a function of segment distance from the root collar is used. The digital 174 175 maize roots in our simulation are around three weeks old while the brace roots develop only after 176 several weeks in a real maize plant; hence, the brace root data are not included in our model.

To compute soil electrical properties, we used Archie's law (Archie, 1942) with an additional term for surface conductivity of the solid phase  $\sigma_{surface}$ , which is assumed to act in parallel (Waxman and Smits, 1968). The relation between soil water content  $\theta$  and  $\sigma_{soil}$  for unsaturated soil is given by Eq. 2, where Archie's fitting parameters (*m* and *d*) vary for different types of soil (Friedman, 2005):

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$$\sigma_{bulk-soil} = \sigma_w n^m S^d + S^{d-1} \sigma_{surface}, \qquad [2]$$

183 where, *S* is the degree of water saturation  $(S = \frac{\theta}{n})$ , *n* the porosity of soil (assumed to be equal to 184 saturated water content:  $\theta_s$ ),  $\sigma_{bulk-soil}$  the bulk electrical conductivity of the soil medium without 185 considering roots (more specifically,  $\sigma_{bulk-loam}$  for loam and  $\sigma_{bulk-sand}$  for sand),  $\sigma_w$  the 186 conductivity of soil fluid phase,  $\sigma_{surface}$  is the surface electrical conductivity of the solid phase of 187 the soil. Sand typically has very low surface conductivity (~10<sup>-5</sup> S/m) while for loam, we assume 188  $\sigma_{surface}$  to be 0.015 S/m (Brovelli and Cassiani, 2011). For Archie's fitting parameters, we use the 189 typical values d = 2 and m = 1.3 (e.g. Werban et al., 2008).  $\sigma_{bulk-soil}$ , in the rhizotron also depends 190 on the electrical conductivity of the nutrient solution  $(\sigma_w)$  in the rhizotron used to grow plants. 191 Measurements from suction cups indicate that  $\sigma_w$  varies between 0.06 to 0.2 S/m (Jougnot et al., 192 2012). We assume  $\sigma_w$  to be 0.2 S/m and choose n as 0.35 (sand) and 0.435 (loam), respectively, for calculating and comparing different pedophysical models. In the following sections, we will 193 194 refer  $\sigma_{bulk-soil}$  as the soil bulk electrical conductivity (with  $\sigma_{bulk-loam}$  and  $\sigma_{bulk-sand}$  to specify soil type) when no roots are present and  $\sigma_{bulk}$  will be used for studies or dataset where both roots 195 196 and soil are present.

## 197 **2.3 Electrical modeling in EIDORS**

The ERT forward problem seeks apparent conductivity or voltage data by solving the Poisson's 198 equation with appropriate boundary conditions with a known electrical conductivity distribution. 199 In ERT inverse problems, we aim at reconstructing an estimate of the electrical conductivity 200 201 distribution within the soil-root domain from apparent conductivity or voltage measurements at its boundary or at some discrete locations within the computational domain. The inverse problem finds 202 203 an approximate  $\sigma$ -distribution that minimizes the data misfit between the virtual measurements and the model predictions in a least-square sense in addition to a regularization term. We use the finite-204 205 element based software EIDORS (Adler and Lionheart, 2006) to solve the forward and inverse problems as it offers flexibility in using different meshing software such as NETGEN (Schöberl, 206 207 1997) and gmsh (Geuzaine and Remacle, 2009). The integration of such meshing software allows 208 creating complex finite-element models for electrical conduction in a soil-root medium. The 209 electrical conduction model for the rhizotron is in purely 2-D (x-z plane, y=0). A point electrode 210 model (Hanke et al., 2011) with a total number of 50 electrodes and a dipole-dipole measurement scheme is used to compute the forward response. All the electrodes are located at the boundary of 211 the computational domain with a similar set-up as in Weigand and Kemna (2017). Three different 212 finite-element meshes are used (Fig. 4). To simulate the ERT data set, the root growth simulation 213 model mesh (SMDL), with an explicit representation of the root architecture is used. The ERT 214 forward model mesh (FMDL), which does not contain the root architectural information, is used 215 to compute the data misfit in the ERT inversion, and the ERT inverse mesh, a comparatively coarse 216 mesh is used to compute the Jacobian in the ERT reconstruction. In the SMDL, either a  $\sigma_{root}$  or 217  $\sigma_{bulk-soil}(\theta)$  value is assigned to each element. The maize primary roots in our simulation have a 218

mean thickness (~ 0.05 cm) which is small compared to the dimensions of rhizotron (20x40 cm),
requiring a very high spatial resolution for roots in the SMDL.

221 To generate a root resolved mesh with high spatial resolution, first we created the binary images of root architectures at various ages (day 5, 7, 10, 12, 15, 18, 22, see Fig. 3). In these binary images, 222 we removed extremely fine root hairs and root branches that were below 0.01 cm in thickness, 223 assuming that such roots have negligible effect on the electric potential distribution. The simplified 224 225 root image represent root branches with a mean diameter of 0.05 cm. Second, we convert binary image into a spline function that traces the boundary of the root surface (red lines in Fig. 4 b) using 226 227 the boundary tracing function "bwboundaries" in MATLAB. The spline function representing the root shape was converted into finite element mesh using gmsh software. The root architecture mesh 228 229 created in this manner possesses superior quality in terms of aspect ratio of elements and is computationally efficient. We then solved the electrical forward problem for the generated  $\sigma$ -map 230 yielding virtual ERT data, which is subsequently inverted using EIDORS. The Gauss-Newton 231 (GN) difference inversion algorithm used in this study requires ERT forward data to be taken on 232 two different conductivity distributions, so that the change in conductivity can be estimated. To do 233 234 so, first we generate a forward data set  $(d_1)$  for a homogeneous  $\sigma$ -distribution as the first medium 235  $(\sigma_1 = 1 S/m)$ . Then the second forward data  $d_2$  is computed on a medium  $\sigma_2$ , which is the sum of soil root electrical conductivity and  $\sigma_1$  (that is,  $\sigma_2 = \sigma_1 + \sigma_{bulk}$ ). Using the change in 236 measurement ( $\delta d = d_2 - d_1$ ), change in conductivity ( $\delta \sigma = \sigma_2 - \sigma_1 \sim \sigma_{bulk}$ ) is estimated by GN 237 one step algorithm (Adler et al., 2007). The high value chosen for  $\sigma_1$  (1 S/m) make sure that the 238 239 change in conductivity and change in measurement can be linearly related as required by GN 240 difference algorithm. The maximum conductivity in sand medium ( $\sim 0.04$ ) and in loam medium ( $\sim$ 241 0.07) is within 10 %, when compared to  $\sigma_1$  ensuring linearity between  $\delta\sigma$  and  $\delta d$ . Finally, the inversion is regularized using a Laplacian matrix (smoothing constraint). To simulate noise in data, 242 243  $\delta d$  is added with 1% random noise proportional to each measurement. The FMDL mesh is used to compute the ERT data ( $\delta d$ ) and the data misfit while the inverse mesh is used for the inversion. 244

# 245 **2.4 Computing average and effective electrical properties**

To get an insight on how a rooted soil might differ from bare soil pedophysical model (Eq. 2), we compare bulk electrical conductivity of soil-root medium, at two different scales: 2x2 cm and 20 x 40 cm.

At smaller scale, the block-wise averaged data, denoted by  $\langle \sigma_{bulk} \rangle$  and  $\langle \theta \rangle$  for electrical 249 250 conductivity and water content respectively, are computed from averaging the corresponding data 251 in the simulation model finite element mesh with an averaging block size of 2cm x 2cm (see Fig. 252 7a). Averaging in each block is done by taking the arithmetic and the harmonic averages of 253 conductivity data of all finite elements within each averaging block to get  $\langle \sigma_{bulk} \rangle$ . The arithmetic averages assumes that the soil-root elements in each averaging block are connected in series while 254 255 the harmonic mean assumes the elements to be in parallel. For  $\langle \theta \rangle$ , we computed only the arithmetic mean. In reality we expect, the real  $\langle \sigma_{bulk} \rangle$  to be in between the arithmetic and 256 harmonic averages. The relation between the collection of averaged data points at every averaging 257 block and at all time (day 5 to , i.e.  $\langle \theta \rangle$  vs  $\langle \sigma_{bulk} \rangle$ , will then approximately mimic the impact of 258 roots at a block-scale on  $\sigma_{bulk}$ , when compared to the Archie's law applied in soils only (Eq.2). 259

At larger scale (rhizotron scale, i.e. 20cm x 40 cm), simple mean of arithmetic and harmonic 260 averages over whole domain may not exactly represent bulk property, as we need to account for 261 the complex structural variations of electrical conductivity distributions and heterogeneity in soil 262 263 electrical property. Hence, to compute  $\sigma_{bulk}$ , at the scale of the rhizotron, we solve the Poisson's equation between two plate electrodes at the boundaries with root included (a root segment has its 264 265 own electrical conductivity) and without root. The computation is repeated for two directions: in horizontal ( $\sigma_{bulkx}$ ) and vertical direction ( $\sigma_{bulkz}$ ). We included in our simulations, the plate 266 267 electrodes that cover the entire left and right walls of the rhizotron as well as top and bottom wall of rhizotron and the ratio of injected current to measured voltage in these electrodes with the 268 geometric factor considerations gives  $\sigma_{bulkZ}$  and  $\sigma_{bulkX}$ . 269

270 **3 Results** 

# 271 **3.1 Electrical measurements on Maize root segments**

Figure 5a shows the experimental data of  $\sigma_{root}$  as a function of root age for Maize. We observe a gradual increase in  $\sigma_{root}$  of intact maize root segments, as the segment distance from root collar increased (Fig. 5a). The trend is different in primary and brace roots, where the brace root conductivity increases much more rapidly with increasing distance of the segment from the root collar compared to primary root segments. The  $\sigma_{root}$  also varies with respect to root cross-sectional area. Our measurements indicate that thinner roots have higher  $\sigma_{root}$  compared to thicker roots (Fig. 5b). This could be due to higher water content of younger roots. Since we measured intact 279 root segments, the surface electrical conductivity of endodermis and contact resistance of stele and cortex layers of the root are accounted for in the measurements. The thicker outer layer (cortex) of 280 281 the root is electrically more insulating than water rich younger roots or inner part (stele) as seen in early studies of Anderson et al. (1976). However, our measurements represent the combined 282 resistivity of cortex and stele in an intact form. Age dependent electrical conductivity variations 283 within a given species were earlier studied in poplar roots (Zenone et al., 2008). Fig. 5a shows that 284 285 within the same species, in addition to age, different types of roots (brace and ground roots) can have different electrical properties. However, in the modeling work, we do not consider the 286 287 development of brace roots as the simulated root system in the model is relatively young (3-weeks old). The blue-curve of Fig. 5a represents the data incorporated in our simulations: 0.0154 288 289  $< \sigma_{root} < 0.03 \, [\text{S/m}].$ 

#### 290 **3.2 Virtual root simulation**

Simulations show that the relative SWC distribution patterns depend on the soil type (Figs. 6 a, b). After 22 days, the depletion is higher is the sand rhizotron as  $\theta_s$  is lower. In the loam, the soil is wetter and the contrast in saturation degree between the rooted and unrooted parts of the soil is much bigger.

295 When translated into electrical conductivity maps including the root electrical properties, we see different trends for sand and loam. For sand, we notice that  $\sigma_{root}$  is always larger than  $\sigma_{bulk-sand}$ 296 and the difference between  $\sigma_{root}$  and  $\sigma_{bulk-sand}$  is always positive (Fig. 6c). For loam, however, 297 298 we see that different regions where  $\{\sigma_{root}, \sigma_{bulk-loam}\}$ , contrast changes with time (Fig. 6d). At 299 initial time,  $\sigma_{bulk-loam}$  is larger than  $\sigma_{root}$  but at day 18, we see different regions, where the difference between  $\sigma_{root}$  and  $\sigma_{bulk-loam}$  is either positive, negative or close to zero. Such contrast 300 301 does not manifest in sand. At day 22, in the upper portion of rhizotron,  $\sigma_{root}$  is greater than  $\sigma_{bulk-loam}$  whereas in the lower portion of rhizotron, the roots are masked by  $\sigma_{bulk-loam}$  (see Fig. 302 6d). In real scenarios, i.e in any soil-root system, there potentially exist three regions, where the 303 difference between  $\sigma_{root}$  and  $\sigma_{bulk-soil}$  is either positive, negative or close to zero. In our study, 304 we observe that at low SWC, the mean of  $\sigma_{root}$  is greater than the mean of  $\sigma_{bulk-loam}$  and at high 305 water content, the mean of  $\sigma_{root}$  is lower than the mean of  $\sigma_{bulk-loam}$ , while in sand, the mean of 306  $\sigma_{root}$  is nearly same as the mean of  $\sigma_{bulk-sand}$  (Fig. 6e). Since electric current flow depends on the 307 gradient of  $\sigma$ -distribution, the effect of roots in ERT experiments will be greater where there is 308 higher  $\{\sigma_{root}, \sigma_{bulk-soil}\}$  contrasts and most importantly, it is time dependent (Fig. 6e). In addition, 309

- the density of roots plays a role in terms of  $\{\sigma_{root}, \sigma_{bulk-soil}\}$  contrasts, for instance at day 22, the upper part of the root system is more conductive than  $\sigma_{bulk-loam}$  in the upper part of the rhizotron and also reflects higher root volume than at initial times. Therefore, at later times (Fig. 6d, day 22), the ERT estimate of water content in the upper region could be misleading due to a stronger root
- 314 influence on  $\sigma_{bulk}$ .

# 315 **3.3 Bulk electrical properties**

The block wise averaged electrical conductivity data points lie along the Archie's curve for low 316 root density regions and deviates significantly from the Archie's curve for high root dense regions 317 (Figs. 7 b and d). In sand, we see more difference between arithmetic and harmonic mean with 318 harmonic mean staying closer to the original pedophysical curve than arithmetic mean (Fig. 7b). 319 In loam, however, there is no big difference between arithmetic and harmonic block wise averaged 320 321 data and both of them change the curvature of the pedophysical relation (Fig. 7d). As expected, when root density is high, the  $\langle \theta \rangle$  vs  $\langle \sigma_{bulk} \rangle$  plot significantly deviates from Eq. 2 and always 322 overestimates  $\sigma_{bulk-sand}$ , whereas areas with very low to zero root density lie along  $\sigma_{bulk-soil}(\theta)$ 323 curve (blue dots on  $\sigma_{bulk-soil}(\theta)$  curve for  $\theta > 0.2$  in Fig. 7b and  $\theta > 0.3$  in Fig. 7d). In loam, the 324  $\langle \theta \rangle$  vs  $\langle \sigma_{bulk} \rangle$  points are scattered around the petrophysical relationship with a tendency of both 325 overestimating  $\sigma_{bulk-loam}$  for  $\theta < 0.2$  and underestimating  $\sigma_{bulk-loam}$  for  $\theta > 0.2$  in the root dense 326 region. This illustrates how roots might affect the relationship between SWC and  $\sigma_{bulk}$ . 327

At rhizotron scale, the effective bulk property shows significant anisotropic affect in sand (notice 328 the difference between  $\sigma_{bulkX}$  and  $\sigma_{bulkZ}$  in Fig. 7c). We expect that the dry sand act as a barrier 329 to the electrical current flow, thereby decreasing the  $\sigma_{bulk}$ . The vertical direction has more 330 331 pronounced anisotropy, when compared to horizontal direction ( $\sigma_{bulkX} > \sigma_{bulkZ}$ ), as we see less deviation of  $\sigma_{bulkX}$  from Archie's law when compared to  $\sigma_{bulkZ}$ . This is due to horizontal layering 332 that develops in the electrical conductivity distribution due to root water uptake, which thereby 333 334 affects current more in vertical direction than in horizontal direction. For loam medium, the anisotropic effect is less when compared to sand. We see from the effective bulk properties that the 335 original pedophysical relation (Archie's law) would rather under-estimate the water content in loam 336 where as it would over-estimate in sand (Fig.7 c and e). Computed  $\sigma_{bulkX}$  and  $\sigma_{bulkZ}$  data points 337 338 lie below  $\sigma_{hulk-soil}(\theta)$  in sand medium whereas above in the loam medium. The rhizotron scale, bulk electrical conductivity deviates from Archie's law quite differently when compared to the 339 340 averaged data at smaller scale. This can be understood as the impact of soil heterogeneity playing a bigger role in influencing the bulk property at large scale whereas at centimeter-scale (2cm x
2cm), the root density plays a major role in the deviating the bulk property from bare soil
pedophysical relation (Eq.[2]).

Table 2 gives the computed anisotropy factor,  $AF = \sigma_{bulkX}/\sigma_{bulkZ}$  for cases with and without roots. As we can see from the table, the main anisotropic affect is due to soil heterogeneity and not the root themselves.

### 347 **3.5 ERT Inversion result**

The GN one step inversion was performed on the virtual measurement data set from the forward conductivity distribution with root system included (Figs. 8 a, d) and also without considering the root system. Figures 8b and 8e shows the ERT inversion with root system included sand and loam medium, respectively. As we can see, the inversion works well in recovering most of the important features of soil water depletion, but sometimes we can observe contamination due to the presence of roots (for example day 18 and 22 in Fig. 8b). Note that for sand the presence of roots increased the electrical conductivity while for loam it decreased the electrical conductivity.

Figures 8c and 8f represents the difference in the inversion results of virtual data from forward 355 conductivity distributions with and without root systems. The inversion result with roots is showed 356 357 in Figs 8 (b and e) but the inversion results without considering the root system are not shown here. The inversion of ERT data without considering root segments were realized by inverting the 358 apparent resistance data resulting from the soil water content map only (see Fig. 6 a,b), without 359 considering the root electrical properties. This difference maps in Figs. 8c and 8f represent the 360 impact of the roots on the ERT inverted  $\sigma$ -distributions. In sand, the error in estimating the 361 electrical conductivity corresponding to the water content of Archie's law can be as low as 2% 362 363 when the roots are small and can reach up to 15% when soil becomes dry and roots occupy the 364 whole rhizotron.

By comparing Figs. 8(a,b,c) and 8 (d,e,f), we can immediately see that in loam, the soil is more conductive than root at most time. Roots are like low conductive wires in the loam medium surrounded by highly conductive soil. Since  $\sigma_{bulk-loam}$  dominates the effective properties, the impact of roots is also lower in loam compared to that of sand. At later time (Fig. 8d, day 22), as root water uptake becomes significant, the contrast between  $\sigma_{bulk-loam}$  and  $\sigma_{root}$  reduces making roots indistinguishable from soil. Figures 8c and 8f indicate that the error in the estimation of the conductivity /water content increases with ongoing root growth. While the error pattern is 372 monotonic in sand increasing with root growth, in loam we see different regions of high and low 373 error depending on soil-root contrast. These errors in  $\sigma$ -estimate manifest in the SWC estimated 374 from Archie's law. We denote here, the volumetric average of water content from RSWMS simulation by:  $\theta_1$ , volumetric average of water content from ERT inverted  $\sigma$ -map without the root 375 electrical properties in the ERT forward data by:  $\theta_2$ , and volumetric average of water content from 376 ERT inverted  $\sigma$ -map with the root electrical properties included in the ERT forward data by:  $\theta_3$ . 377 We show  $\theta_1$ ,  $\theta_2$  and  $\theta_3$  as a function of time in Figure 9 (a, d). The difference between  $\theta_1$  and 378  $\theta_2$  is the error induced due ERT inversion procedure alone while the difference between  $\theta_1$  and 379  $\theta_3$  is the error induced due to ERT method as well as the root segments. In Figure 9 (b,c,e,f), we 380 show that these errors in absolute and relative terms are more pronounced when the root system is 381 large. When the root is young (age <10 days), the absolute error between  $\theta_1$  and  $\theta_2$  is same as the 382 absolute error between  $\theta_1$  and  $\theta_3$  indicating that root segments has no significant impact in water 383 content estimates (Fig. 9b and e). 384

## 385 4 Discussions

386 Soil-root water flow modeling together with root electrical measurements reveals that soil-root 387 electrical conductivity contrasts changes over time (Fig. 6) as a function of soil type and root water uptake. At centimeter-scale (2cm x 2 cm), the root play a major role in deviating  $\sigma_{bulk}$  from 388 Archie's law. Block-wise averaged data ( $\langle \theta \rangle$  vs  $\langle \sigma_{bulk} \rangle$ ) shows that rooted soil deviates in terms 389 of pedophysical relation from bare soil, where there is higher root density (Fig.7). This is consistent 390 with the experimental observation made by Michot et al., (2016), where they found that bare-soil 391 pedophysical relation is inadequate to explain  $\sigma_{bulk}(\theta)$  in the rooted zone. At decimeter scale, the 392  $\sigma_{bulk}$  computed using plate electrode reveals anisotropy and different behavior as compared to the 393 centimeter scale averaged data (see Fig. 7 b and c). We also observe an anisotropy factor of around 394 395 six for fully mature root systems. This is mostly due to water content distribution pattern induced by root water uptake. At rhizotron scale, anisotropy is stronger in sand, when compared to loam, 396 397 and increases non-linearly with root growth (Table 2).

The modeling results clearly show that roots impacts ERT results. The degree of impact further depends on electrical conductivity contrast between root and soil. To characterize the specific impact of roots in ERT monitored water content estimates, we need the knowledge of electrical conductivity contrast between root and soil as a function of space and time. Estimating this contrast between root and soil, however, is not so straightforward and difficult, as they are root type, root 403 age, root radius, soil type and water content dependent. Although the maize simulations in this 404 study indicates that water content is the dominant factor affecting bulk electrical conductivity, other 405 factors do play a role including the root connectivity that induces electrical anisotropy. Further 406 upscaling the electrical properties derived from centimeter scale (root segment) to decimeter scale 407 (rhizotron) to field scale (~100 meter) is very important to develop a proper pedophysical relation 408 that completely eliminates the root impact in the water content estimate.

The error shown in Figure 9 represents the impact of root, when soil is completely characterized by a known pedophysical function such as Archie's law or Waxman-smit model. However, in some situations, a proper pedophysical function that transforms electrical conductivity of a soil to soil water content is not known a priori and hence in addition to roots, error in water content estimates can be introduced due to usage of wrong pedophysical model.

414 Since our model indicates a non-negligible anisotropy factor in the electrical conductivity, ERT injection scheme should consider exploiting anisotropy to retrieve better information, for example, 415 by having an injection scheme that maximizes the sensitivity in the region of anisotropy. Since 416 anisotropy in  $\sigma$  changes with development of the root system in soil, one could also have time-417 418 dependent ERT injection schemes for the time-lapse ERT. A prior knowledge of time dependent electrical conductivity contrasts between soil and root, for a given crop, can definitely help in 419 420 designing optimized ERT injection scheme for the future field experiments. The volumetric total 421 water content shows a larger error for sandy soil (Fig.9). However, yet the overall trend of decrease 422 in total water content due to root water uptake is recovered. The difference ERT inversion algorithm works well in recovering the overall structure of water uptake. For maize roots, the water 423 424 uptake process dominates the  $\sigma$ -distribution of the soil-root system as reconstructed with ERT. It 425 is also worth noting that there are various other root architectures such as tap root systems, which 426 still need further investigations on their electrical anisotropy at rhizotron scale and at field scale.

Finally, we considered a very limited range in  $\sigma_{root} - \sigma_{bulk-soil}$  variability. In reality, the range of variations in  $\sigma_{root} - \sigma_{bulk-soil}$  could differ depending on the type of roots and the value of  $\sigma_w$ (Fig.1). As in agricultural fields, even in two-dimensional rhizotron experiments, air filled cracks can manifest in the soil, potentially influencing ERT measurements. In our model, we did not consider such real-world phenomena and limited our study only to the impact of roots. We also ignored rhizosphere processes such as root exudation, which could affect the water content

estimates. In reality, soil-root systems are three-dimensional structures and two-dimensional 433 rhizotron approximations may not represent an accurate model for three-dimensional electrical 434 435 conductivity in real soil-root environments (e.g. cropped fields). We also ignored the anisotropy of  $\sigma$  inside the root structure (stele-cortex variations), which may have a considerable effect on ERT 436 437 measurements. Such structural variations may induce even higher degree of anisotropy in the electrical conductivity. Our next step will be the validation of our findings in real experiments and 438 439 under even more realistic conditions, accounting, amongst other aspects, for the specific rhizosphere properties, and to extend the studies to include complex conductivity (induced 440 441 polarization) properties.

#### 442 **5.** Conclusions

We simulated an electrical conductivity model of a soil-root continuum in the rhizotron geometry. The roots were explicitly represented in the  $\sigma$ -distribution and root water uptake was simulated using mechanistic water flow models in soils and roots. We designed experiments on intact root segments to measure electrical properties of roots ( $\sigma_{root}$ ). Our measurements on maize root segments indicated that  $\sigma_{root}$  is a function of distance from the root collar and root type (primary and brace roots). We incorporated the distance variations of primary roots into our model based on a polynomial fit.

450 The GN type ERT inversion results (Figs. 8) reveal that exclusion of the explicit representation of

451 roots in the forward model results in an error of 5 to 15% in  $\sigma$ . Even though the effect of roots at

452 rhizotron scale is not evident in the bulk property analysis of conductivity data (table 2), it is evident

453 in ERT inversion result. This indicates the importance of incorporating the effect of roots in the

454 pedophysical model.

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	$\theta r [\mathrm{cm}^3\mathrm{cm}^{-3}]$	$\theta s [\text{cm}^3 \text{ cm}^{-3}]$	<i>a</i> [1/cm]	n	Ks [cm/day]	λ
Sand	0	0.35	0.05	2	100.24	0.5
Loam	0.078	0.435	0.036	1.56	25	0.6

Table 1: Soil hydraulic properties.  $\theta r$ : Residual water content,  $\theta s$ : Saturated water content, a, nand  $\lambda$ : shape parameters in van Genuchten-Mualem equations, *Ks*: saturated soil hydraulic conductivity.

a) Sand without	Time:	Day 5	Day 10	Day 15	Day 18	Day 22
roots	$\sigma_{bulkZ}$	0.0127	0.0074	0.0041	0.0015	0.0002
	[S/m]					
	$\sigma_{bulkX}$	0.0215	0.0144	0.0106	0.0064	0.0012
	[S/m]					
	AF	1.68	1.93	2.58	4.26	6.24
b) Sand with roots	$\sigma_{bulkZ}$	0.0128	0.0077	0.0045	0.0018	0.0002
	[S/m]					
	$\sigma_{bulkX}$	0.0215	0.0144	0.0108	0.0066	0.0012
	[S/m]					
	AF	1.67	1.88	2.39	3.73	6
c) Loam without	$\sigma_{bulkZ}$	0.0568	0.0449	0.0370	0.0279	0.0166
roots	[S/m]					
	$\sigma_{bulkX}$	0.0594	0.0482	0.0417	0.0337	0.0190
	[S/m]					
	AF	1.04	1.07	1.12	1.19	1.14
	$\sigma_{bulkZ}$	0.0566	0.0447	0.0369	0.0281	0.0170
d) Loam with	[S/m]					
roots	$\sigma_{bulkX}$	0.0593	0.0481	0.0414	0.0334	0.0194
	[S/m]					

AF	1.04	1.07	1.12	1.19	1.14

599 Table 2: Effective electrical conductivity in [S/m] and anisotropy factor at rhizotron scale 600 computed using simulated plate electrodes at boundaries.

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604 Figures:





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Figure 1. Comparison of soil and root electrical conductivity. The envelops of  $\sigma_{soil}$  (some with and some without roots) and  $\sigma_{root}$  are shown as shaded areas.

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Figure 2: Flow chart for the simulation of Virtual Rhizotron drying experiment. First, a simulation of root water uptake and root growth of a maize plant in a rhizotron is run with a soil-plant water flow model (RSWMS, Javaux et al., 2008), which generates maps of soil water distribution ( $\theta$ ) and of root architecture evolution. Then these distributions are transformed into detailed electrical conductivity ( $\sigma$ ) maps through bio/pedo-physical relations. Third, these distributions are used to simulate a virtual ERT measurement and inversion scheme to get a coarser distribution of  $\sigma$ estimates (see text for further details).





621 Figure 3: root architectural evolution shown at different times.



Figure 4: a) Virtual rhizotron schematic, b) binary image of schematic root architecture used to
generate mesh. The red region represents the spline curve that envelops the root surface. c)
Simulation model (SMDL): simulation mesh with explicit root architecture and schematic
conductivity distribution map, d) forward model (FMDL): forward and inverse mesh, and e) regular
uniform grid used to simulate Richards' equation.

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Figure 5: Measurement data on Maize roots a)  $\sigma_{root}$  vs distance from root collar, b)  $\sigma_{root}$  vs root cross sectional area. The quadratic fit is shown as solid line while measurement data is represented at discrete locations as circles (primary root) and crosses (brace root). The blue curve in Figure 5a is the data used in simulation model.





636Figure 6: Volumetric water saturation distribution in a) sand and b) loam, and its corresponding σ-637maps in c) sand and d) loam, e) variability of  $\sigma$  in the rhizotron at different times. The vertical bars638at various times represent the minimum and maximum value of  $\sigma$ , respectively.



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Figure 7: a) A section of the SMDL mesh with averaging blocks shown in red squares. The root 640 elements are in brown color, b) Comparison of Archie's law with block-wise arithmetic averaged 641 quantities ( $\langle \sigma_{bulk} \rangle$  vs.  $\langle \theta \rangle$ ) in sand, c) Comparison of Archie's law with block-wise harmonic 642 averaged quantities ( $<\sigma$  bulk> vs.  $<\theta>$ ) in sand, d) Comparison of Archie's law with rhizotron 643 scale effective bulk property in sand, e) Comparison of Archie's law with block-wise arithmetic 644 645 averaged quantities ( $\langle \sigma \rangle$  bulk vs.  $\langle \theta \rangle$ ) in loam, f) Comparison of Archie's law with block-wise 646 harmonic averaged quantities ( $<\sigma$  bulk> vs.  $<\theta>$ ) in loam, g) Comparison of Archie's law with 647 rhizotron scale effective bulk property in loam.



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Figure 8: Sand: a) detailed electrical conductivity map of maize root at different times; b)
tomography inversion with the root conductivity included in the forward model; c) difference
between the inversions results with and without root conductivity accounted for in the forward
model. Green circles represent the electrode positions. Loam: d) Conductivity map of maize root
at different time; e) tomography inversion with the root conductivity included in the forward
model; f) difference between the inversions results with and without root conductivity in the



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Figure 9: a) Comparison of normalized volume averaged water content, obtained from simulated SWC (denoted as  $\theta_1$ ) and ERT imaging without and with inclusion of  $\sigma_{root}$  denoted by  $\theta_2$  and  $\theta_3$ , respectively as a function of different root growth time in sand, b) Absolute error between  $(\theta_1, \theta_3)$ and  $(\theta_1, \theta_2)$ , c) relative error between  $(\theta_1, \theta_3)$  and  $(\theta_1, \theta_2)$ . Figures 9 (d,e,f) same as 9 (a,b,c) but in loam medium.