

1 **Effective hydraulic properties of 3D virtual stony soils identified  
2 by inverse modeling**

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9

10 **Core Ideas**

11     ▪ Virtual stony soils with different rock fragment contents were generated in 3D using the Hydrus 2D/3D  
12       software.

13     ▪ Evaporation experiments and unit-gradient experiments were numerically simulated.

14     ▪ We used inverse modelling with the Richards equation to identify effective hydraulic properties of virtual  
15       stony soils.

16     ▪ The identified hydraulic properties were used to evaluate the scaling models of calculating hydraulic  
17       properties of stony soils.

18

19 **Keywords**

20 Soil hydraulic properties, water retention curve, hydraulic conductivity, stony soil, inverse modeling, Hydrus 2D/3D

21 **Abstract**

22 Stony soils that have a considerable amount of rock fragments (RF) are widespread around the world. However,  
23 experiments to determine effective hydraulic properties of stony soils (SHP), i.e. the water retention curve (WRC) and  
24 hydraulic conductivity curve (HCC), are challenging. Installation of measurement devices and sensors in these soils  
25 is difficult and the data are less reliable because of their high local heterogeneity. Therefore, effective properties of  
26 stony soils especially in unsaturated hydraulic conditions are still not well understood. An alternative approach to

27 evaluate the SHP of these systems with internal structural heterogeneity is numerical simulation. We used the Hydrus  
28 2D/3D software to create virtual stony soils in 3D and simulate water flow for different volumetric fractions of RF,  $f$ .  
29 Stony soils with different values of  $f$  from 11 to 37 % were created by placing impermeable spheres as rock fragments  
30 in a sandy loam soil. Time series of local pressure heads in various depths, mean water contents, and fluxes across the  
31 upper boundary were generated in a virtual evaporation experiment. Additionally, a multi-step unit gradient simulation  
32 was applied to determine effective values of hydraulic conductivity near saturation up to  $pF = 2$ . The generated data  
33 were evaluated by inverse modeling, assuming a homogeneous system, and the effective hydraulic properties were  
34 identified. The effective properties were compared with predictions from available scaling models of SHP for different  
35 values of  $f$ . Our results showed that scaling the WRC of the background soil based on only the value of  $f$  gives  
36 acceptable results in the case of impermeable RF. However, the reduction of conductivity could not be simply scaled  
37 by the value of  $f$ . Predictions were highly improved by applying the Novák, Maxwell, and GEM models to scale the  
38 HCC. The Maxwell model matched the numerically identified HCC best.

39 **1. Introduction**

40 Stony soils are soils with a considerable amount of rock fragments (RF) and are widespread in mountainous and  
41 forested watersheds around the world (Ballabio et al., 2016; Novák and Hlaváčiková, 2019). RF in soil are particles  
42 with an effective diameter of larger than 2 mm (Tetegan et al., 2015; Zhang et. al., 2016). Their existence in soil  
43 influences the two constitutive soil water relationships known as soil hydraulic properties (SHP) i.e. water retention  
44 curve (WRC), and hydraulic conductivity curve (HCC) (Russo, 1988; Durner and Flühler, 2006). The accurate  
45 identification of SHP is a prerequisite for adequate prediction of water flow in soil with the Richards equation  
46 (Farthing and Ogden, 2017; Haghverdi et al., 2018). The SHP depend on soil texture and structure (Kutilek, 2004;  
47 Lehmann et al., 2020), and are influenced by the presence of RF in soil. It is generally accepted that RF decrease the  
48 water storage capacity of soils and its effective unsaturated hydraulic conductivity. In contrast, the formation of  
49 macropores in the vicinity of embedded RF may lead to an increase in saturated hydraulic conductivity. While  
50 experimental evidence and theoretical analyses show that the volumetric fraction of RF,  $f$  (v/v), has the highest  
51 influence on effective SHP of stony soil, the effect of other characteristics of RF such as their porosity, shape, size,  
52 arrangement, and orientation towards flow is less clear (Hlaváčiková and Novák, 2014; Hlaváčiková et al., 2016;  
53 Naseri et al., 2020). Up to the present, two approaches have been dominant in identifying the hydraulic behavior of  
54 stony soils: I) Experimental setups with the aim of measuring SHP of stony soils in the field or in controlled systems

55 in the laboratory (Cousin et al., 2003; Dann et al., 2009; Grath et al., 2015; Beckers et al., 2016, Naseri et al., 2019),  
56 and II) Development of empirical, physical or physico-empirical approaches to scale hydraulic properties of  
57 background soil based on the value of  $f$  and characteristics of RF (Novák et al., 2011; Naseri et al., 2020). These two  
58 approaches have some systematic limitations that restrict their applications in investigating the hydraulic behavior of  
59 stony soils. Installation of sensors and measurement instruments in the stony soils are technically demanding (Cousin  
60 et al., 2003; Verbist et al., 2013, Coppola et al., 2013; Stevenson et al., 2021), undisturbed sampling is laborious  
61 (Ponder and Alley, 1997), relatively larger samples are required (Germer and Braun, 2015), and the measured data  
62 might be more inconsistent due to the higher local heterogeneity of such soils (Baetens et al., 2009; Corwin and Lesch,  
63 2005). Furthermore, some of the available scaling models to obtain effective SHP are conceptually oversimplified and  
64 they exclusively consider the value of  $f$  as the only input parameter (Bouwer and Rice, 1984; Ravina and Magier,  
65 1984). Additionally, they assume impermeable RF and are proposed mainly for saturated flow conditions. These  
66 scaling models need a systematic verification under variably-saturated conditions using experimental data or 3D  
67 simulations. Some reviews of these models and their evaluation are available in the literature (Brakensiek et al., 1986;  
68 Novák et al., 2011; Beckers et al., 2016; Naseri et al., 2019).

69 Hlaváčiková and Novák (2014) proposed a model to scale the HCC of the background soil, parametrized with the van  
70 Genuchten–Mualem (van Genuchten, 1980) model, using the model of Bower and Rice (1984). Hlaváčiková et al.,  
71 (2018) used the water content of RF as input parameter to scale the WRC of the background soil. Naseri et al. (2019)  
72 used the simplified evaporation method (Peters et al., 2015) to determine experimentally the effective SHP of small  
73 soil samples containing various amounts of RF. Their study criticizes the application of the scaling models developed  
74 for saturated stony soils to unsaturated conditions and emphasizes the need to develop approaches that consider more  
75 characteristics of the RF to calculate SHP of the stony soils.

76 Recent advancements in computational hydrology and computing power suggest the numerical simulation of soil  
77 water dynamics as a promising alternative to the measurement of effective SHP of heterogeneous soils (Durner et al.,  
78 2008; Lai and Ren, 2016; Radcliffe and Šimůnek, 2018). Numerical simulations have several advantages. They do not  
79 demand strict experimental setups, are repeatable under a variety of initial and boundary conditions, and in contrast  
80 to the laboratory experiments, space and time scales are not restrictive factors in the simulations. These assets have  
81 made them a favorable tool in water and solute transport modeling in heterogeneous soils (Abbasi et al., 2003;

82 Šimůnek et al., 2016). However, with few exceptions, heterogeneous soils like stony soils have been simulated only  
83 for simplified cases, i.e., either under fully saturated conditions or with reduced dimensionality, i.e., simulations of  
84 stony soils in two spatial dimensions (2D). Novák et al. (2011) calculated effective saturated hydraulic conductivity  
85 ( $K_s$ ) of soils containing impermeable RF using steady-state simulations with the software Hydrus 2D which solves the  
86 Richards equation in two spatial dimensions. They derived a linear relationship between  $K_s$  of stony soils and  $f$ .  
87 Hlaváčiková et al. (2016) simulated different shapes and orientations of RF in Hydrus 2D to obtain the effective  $K_s$  of  
88 the virtual stony soils. Beckers et al. (2016) used Hydrus 2D simulations to extend the investigations towards the  
89 impact of  $f$ , shape, and size of RF on the HCC. They also identified effective SHP of a clay stony soil using laboratory  
90 evaporation experiments for  $f$  values up to 20 % (v/v).

91 The inverse modeling approach has been applied to identify effective hydraulic properties of soils in laboratory  
92 experiments (Ciollaro and Romano, 1995; Hopmans et al., 2002; Nasta et al., 2011), in lysimeters and field (Abbaspour  
93 et al., 1999; Abbaspour et al., 2000), virtual lysimeters with internal textural heterogeneity (Durner et al., 2008; Schelle  
94 et al., 2013), and WRC of stony soils through field infiltration experiments (Baetens et al., 2009). Although theoretical  
95 studies and laboratory investigations on packed samples are insufficient to understand fully the hydraulic processes in  
96 stony soils, they do lead the way to the improvement and validation of effective models and their application at the  
97 field and even larger scales. Inverse modeling is arguably the best approach to achieve these aims because it allows to  
98 validate effective models using process modeling. Our aim in this study was to investigate the application of inverse  
99 modeling to identify the effective SHP of 3D virtual stony soils and to explore its applicability to these soil systems  
100 as an example of internal structural heterogeneity. We were interested in answering the following questions:

101 i) Is it possible to describe the dynamics in the heterogeneous 3D system with the 1D Richards equation assuming a  
102 homogeneous soil?

103 ii) If so, what are the effective SHP of stony soils and how are they related to the SHP of the background soil?

104 To answer these questions we conducted forward simulations of water movement in 3D using the Richards equation  
105 as variably-saturated flow model. We created stony soils by embedding voids representing impermeable spherical RF  
106 as inclusions into a homogeneous background soil. Then we simulated transient evaporation experiments and stepwise  
107 steady-state, unit-gradient infiltration experiments in 3D. The generated data were used as an input to a 1D inverse

108 model to obtain the effective SHP of stony soils, and these properties were used to evaluate and compare the available  
109 scaling models of SHP for stony soils.

110 **2. Materials and methods**

111 **2.1. Simulation model**

112 The Hydrus 2D/3D software was used to generate virtual stony soils and simulate the water flow in the created three-  
113 dimensional geometries. Water flow in Hydrus 2D/3D is modelled by the Richards equation (Šimůnek et al., 2006;  
114 2008), which is the standard model for variably-saturated water flow in porous media. The Hydrus 2D/3D software  
115 solves the mixed form of the Richards equation numerically using the finite-element method and an implicit scheme  
116 in time (Celia et al., 1990; Šimůnek et al., 2008; 2016; Radcliffe and Šimůnek, 2018). The three-dimensional form of  
117 the Richards equation under isothermal conditions, without sinks/sources, and assuming an isotropic hydraulic  
118 conductivity is:

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

119 where  $\theta$  is the volumetric water content ( $\text{cm}^3 \text{ cm}^{-3}$ ),  $t$  is time (s),  $h$  is the pressure head (cm), and  $K(h)$  is the hydraulic  
120 conductivity function ( $\text{cm d}^{-1}$ ).  $x$ , and  $y$  (cm) are the horizontal Cartesian coordinates, and  $z$  (cm) is the vertical  
121 coordinate, positive upwards. We used the van Genuchten-Mualem model to parametrize the WRC and HCC (van  
122 Genuchten, 1980):

$$S_e(h) = \frac{\theta(h) - \theta_r}{\theta_s - \theta_r} = [1 + (\alpha h)^n]^{-m} \quad (2)$$

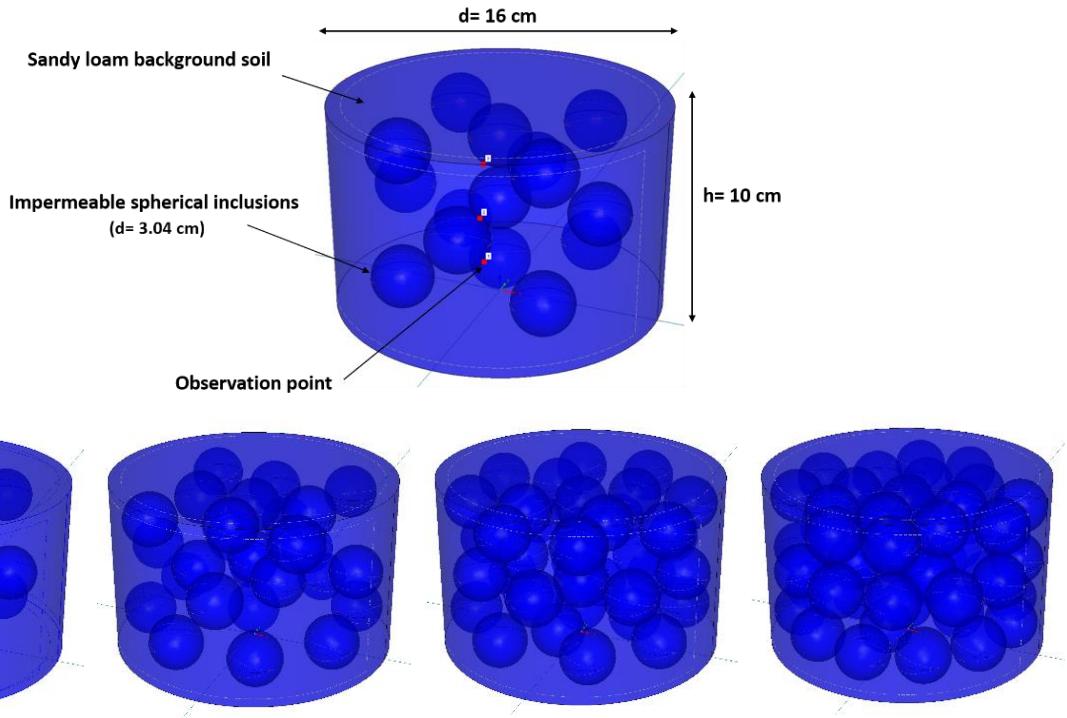
123 and

$$K(h) = K_s S_e^\tau \left[ 1 - \left( 1 - S_e^{\frac{1}{m}} \right)^m \right]^2 \quad (3)$$

124 where  $\theta_s$  and  $\theta_r$  are the saturated and residual water contents ( $\text{cm}^3 \text{ cm}^{-3}$ ), respectively,  $S_e(h)$  is the effective saturation  
125 (-),  $\alpha$  ( $\text{cm}^{-1}$ ) is a shape parameter,  $n$  is an empirical parameter related to the pore size distribution (-) and  $m = 1 -$   
126  $1/n$ ,  $K_s$  is the saturated hydraulic conductivity and  $\tau$  is a tortuosity/connectivity parameter (-).

127 **2.2. 3D geometries representing stony soils**

128 The virtual stony soils in 3D were created by placing spherical inclusions in a background soil. In accordance with  
129 real laboratory experiments (not reported here), we generated virtual soil columns as cylinders with a diameter of 16  
130 cm and a height of 10 cm and an total volume of  $\approx 2011 \text{ cm}^3$ . The inclusions were considered as voids representing  
131 impermeable RF embedded in the background soil. Configurations and characteristics of the created 3D geometries  
132 of stony soils are illustrated in Fig. 1. Each spherical inclusion had a diameter of 3.04 cm and a volume of  $\approx 14.7 \text{ cm}^3$ .  
133 Stony soils with different values of  $f$  were created by including different numbers of spherical inclusions in the soil  
134 column. A total number of 15, 27, 39, and 51 spherical inclusions in each column led to four volumetric rock fragment  
135 contents of 11.0, 19.8, 28.5 and 37.3 % (v/v). Spheres were arranged in the column in three layers. The spheres'  
136 centers were in depths of 2.5, 5.0 and 7.5 in the column and each layer was packed with one-third of the total number  
137 of intended spheres. Furthermore, observation points at selected nodes of the numerical grid were inserted in each of  
138 the three depths of the column (i.e. 2.5, 5.0 and 7.5 cm) in the background soil and not in close vicinity of the inclusions  
139 to provide time series of soil water pressure head for the inverse simulations. For the background soil, a homogenous  
140 sandy loam soil was considered with the van Genuchten-Mualem model parameters  $\theta_s = 0.410 \text{ (cm}^3 \text{ cm}^{-3}\text{)}$ ,  $\theta_r =$   
141  $0.065 \text{ (cm}^3 \text{ cm}^{-3}\text{)}$ ,  $\alpha = 0.01 \text{ (cm}^{-1}\text{)}$ ,  $n = 2.0 \text{ (-)}$ ,  $\tau = 0.5 \text{ (-)}$ , and  $K_s = 100 \text{ (cm d}^{-1}\text{)}$ . The targeted mesh size for the  
142 different simulations was set to 0.25 cm. The dependency of the numerical solution on the mesh size was tested with  
143 some refined meshes and negligible differences in the results were obtained for different mesh sizes.



144

145

*Figure 1: Visualization of the generated stony soils in 3D, including the dimension of rock fragments and soil cylinder and location of the observation points (top). Bottom row shows rock fragment contents of 11.0, 19.8, 28.5 and 37.3 %, from left to right.*

146

### 147 **2.3. Forward simulations**

148 We simulated evaporation (EVA) (Peters and Durner, 2008) and multistep unit gradient (MSUG) experiments (Sarkar  
 149 et al., 2019). For EVA, a linear distribution of pressure head (-2.5 cm top, +7.5 cm bottom) was used as the initial  
 150 condition. The boundary conditions were no-flux at the bottom and atmospheric with a constant potential evaporation  
 151 rate of 0.6 ( $\text{cm d}^{-1}$ ) and zero precipitation at the top. The EVA experiments were simulated for 10 days and the time  
 152 series of pressure heads at each observation point, the initial volumetric water content and the cumulative evaporation  
 153 and evaporation rate were collected for later use in the inverse simulations.

154 In the virtual MSUG experiment, the soil column was initially fully saturated with a constant pressure head of 0 cm.  
 155 A sequence of step-wise decreasing constant pressure heads was assigned to the upper and lower boundaries of the  
 156 column. The duration of the virtual MSUG experiment was 100 days and the pressure head in the upper and lower  
 157 boundaries was simultaneously decreased stepwise to a pressure head of -100 cm. The applied pressure heads  $h_i$  were  
 158 0, -1, -3, -10, -20, -30, -60, and -100 cm, respectively. Time steps were chosen such that a steady-state flow condition  
 159 was reached for each pressure step, indicated by identical water fluxes at the top (inflow) and bottom (outflow)

160 boundaries and constant pressure heads at the observation points. The hydraulic conductivities at the respective  
161 pressure heads  $h_i$  data were calculated by dividing the steady-state water flux rates ( $\text{cm}^3 \text{ d}^{-1}$ ) by the total surficial area  
162 of the soil column ( $\approx 202 \text{ cm}^2$ ).

163 The converging and diverging flow field around obstacles produces even under unit gradient conditions spatially  
164 different pressure heads, and as opposed to saturated conditions, these different pressure heads are under unsaturated  
165 conditions associated with different water saturations and different local hydraulic conductivities. We were interested  
166 in whether and to what extent this could lead to nonlinear effects in the derivation of the effective hydraulic properties,  
167 in particular the effective HCC. Furthermore, since the flow field for a given volume fraction of obstacles depends on  
168 dimensionality, i.e., is different in a 2D simulation than in a 3D simulation, studying the effects in the unsaturated  
169 region was one of the main motivations for performing this numerical analysis in 3D.

170 **2.4. Inverse modeling of evaporation in 1D**

171 A 10-day EVA experiment in 1D was simulated with the software package HYDRUS-1D (Šimůnek et al., 2006; 2008)  
172 to obtain the SHP parameters using inverse modeling. The generated data from the EVA and MSUG forward  
173 simulations in 3D were used as an input to the 1D inverse simulations. Time series of the pressure heads at three  
174 observation depths, mean volumetric water contents in the column during the virtual EVA experiment, and the data  
175 points of the effective HCC from the virtual MSUG experiment were used as data in the objective function. The time  
176 series of the mean volumetric water content was calculated from the initial water content, cumulative evaporation and  
177 soil volume. The measurement range for pressure heads used in the objective function was from saturation down to -  
178 2000 cm. This reflects a setup with laboratory tensiometers with boiling delay (Schindler et al., 2010). The time series  
179 of the simulated evaporation rates from the 3D simulations were used as the time variable atmospheric boundary  
180 condition for the 1D inverse simulations. The 1D soil profile was 10 cm long and was discretized into 100 equally  
181 sized finite elements. Similar to the 3D simulations, three observation points were defined in the depths of 2.5, 5.0  
182 and 7.5 cm. A no-flux boundary condition was used at the bottom. The six parameters of the van Genuchten model  
183 occurring in Eq. (2) and (3) were all simultaneously estimated by inverse modeling. The weighted-least-squares  
184 objective function was minimized by the SCE-UA algorithm (Duan et al., 1992). The data obtained from the EVA  
185 experiment allow to identify the WRC from saturation to the pressure where the tensiometers fail, and the HCC in the  
186 mid to dry range of the SHP (roughly -100 to -2000 cm pressure head), while the MSUG provides a precise

187 determination of the HCC in the wet range (Sarkar et al., 2019; Durner and Iden, 2011). The EVA experiments do not  
188 provide information on hydraulic conductivity near water-saturation (Peters et al., 2015). Therefore, we included the  
189 obtained data from the virtual MSUG experiment in the object function for the inverse simulation of the virtual EVA  
190 experiments, to improve the uniqueness of the inverse solution and the precision of the identified HCC near saturation  
191 (see Schelle et al., 2010, for another example).

192 **2.5. Predicting SHP of virtual stony soils by scaling models**

193 The SHP of stony soils obtained by inverse modeling were compared to SHP that are predicted by available scaling  
194 models and used for their evaluation. Considering that  $f$  has the dominant influence on the WRC of a stony soil, a  
195 common approach is partitioning the WRC and HCC of stony soil based on the volume of each component in the soil-  
196 rock mixture and calculating the effective SHP of stony soil using the volume averaging or the composite-porosity  
197 model. The general form of the WRC model considers the moisture contents of the background soil  $\theta_{\text{soil}}(h)$  ( $\text{cm}^3 \text{cm}^{-3}$ )  
198 and embedded rock fragments  $\theta_{\text{rock}}(h)$  ( $\text{cm}^3 \text{cm}^{-3}$ ) to calculate the effective WRC of stony soils  $\theta_m(h)$  ( $\text{cm}^3 \text{cm}^{-3}$ )  
199 (Flint and Childs, 1984; Peters and Klavetter, 1988) with the following form in the full moisture range (Naseri et al.,  
200 2019):

$$\theta_m(h) = f\theta_{\text{rock}} + (1 - f)\theta_{\text{soil}} \quad (4)$$

201 A typical assumption in stony soils hydrology is that the porosity of RF is negligible. In this case, Eq. (4) reduces to  
202 (Bouwer and Rice, 1984):

$$\theta_m(h) = (1 - f)\theta_{\text{soil}} \quad (5)$$

203 For the effective hydraulic conductivity of stony soils, some scaling models are developed for saturated conditions  
204 that might apply to the hydraulic conductivity at any pressure heads. The simplest scaling model accounts only for the  
205 reduction in the cross-sectional area available for flow of water. This leads to the equation (Ravina and Magier, 1984):

$$K_r = 1 - f \quad (6)$$

206 where  $K_r$  (-) is the relative hydraulic conductivity of stony soil, i.e.,  $K_r = K_m/K_{\text{soil}}$ , where  $K_m$  is the effective hydraulic  
207 conductivity of the stony soil ( $\text{cm d}^{-1}$ ), and  $K_{\text{soil}}$  is the conductivity of the background soil ( $\text{cm d}^{-1}$ ).

208 In a more recent approach, Novák et al. (2011) developed a linear relationship based on the 2D numerical simulation  
209 results as a first approximation to scale the saturated hydraulic conductivity of stony soils:

$$K_r = 1 - \alpha f \quad (7)$$

210 The parameter  $\alpha$  was reported to depend on the texture of the background soil, with a range between 1.1 for sandy  
211 clay to 1.32 for clay. This model is easy to apply, but it requires the estimation of the parameter  $\alpha$  to calculate  $K_r$ . For  
212 our calculations, we assumed  $\alpha = 1.2$  for the sandy loam background soil used in our study.

213 Another model that has been developed for mixtures with spherical inclusions is the Maxwell model (Maxwell, 1873;  
214 Corring and Churchill, 1961; Peck and Watson, 1979; Zimmermann and Bodvarsson, 1995). It takes the value of  $f$ ,  
215 hydraulic conductivity of the background soil and hydraulic conductivity of inclusions into account to calculate the  
216 hydraulic conductivity of the stony soil. In the special case of impermeable inclusions, Maxwell model reduces to:

$$K_r = \frac{2(1-f)}{2+f} \quad (8)$$

217 A recently developed model by Naseri et al. (2020), which is based on the general effective medium theory (GEM),  
218 allows considering effects of permeability, shape, and orientation of RF on the effective HCC. For impermeable RF,  
219 the GEM model reduces to the following form:

$$K_r = \left(1 - \frac{f}{f_c}\right)^t \quad (9)$$

220 where  $f_c$  is the critical  $f$  with values between 0.84 and nearly 1 and  $t$  is a shape parameter with values between 1.26  
221 and nearly 1.5 for spherical RF. In this study, we set  $f_c = 0.982$  (v/v) according to the size ratio of the RF to the  
222 background soil, and  $t = 1.473$  for spherical RF (for details, see appendix in Naseri et al., 2020).

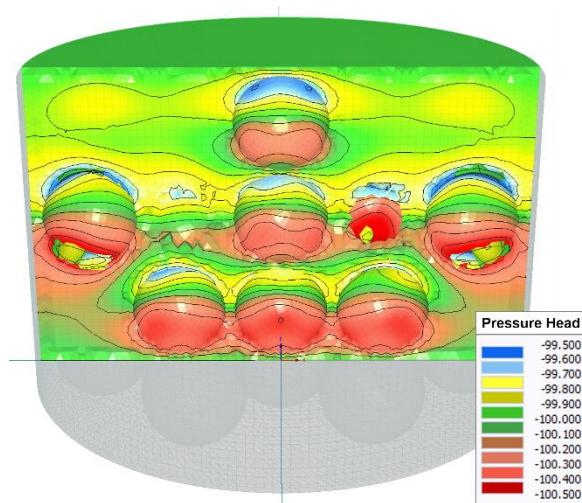
223 It should be noted that all approaches apply at any pressure head  $h_i$ , i.e., the scaling that is originally developed for  
224 saturated conditions with locally constant hydraulic conductivity in the background soil is equally applied to  
225 unsaturated conditions.

226 **3. Results and discussion**

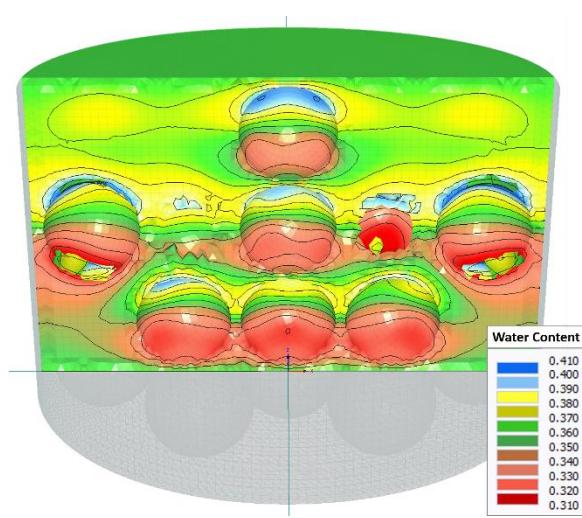
227 **3.1. Flow field and variability of state variables in the virtual MSUG experiment**

228 Figure 2 visualizes the pressure head (cm), water content ( $\text{cm}^3 \text{ cm}^{-3}$ ), and velocity ( $\text{cm d}^{-1}$ ) in a 2D cross section in the  
229 center of the soil column through the forward simulation of the MSUG experiment. The profile is shown for the steady  
230 state flux situation with a pressure head of -100 cm and the stony soil with  $f = 28.5 \%$ . The Figure shows a considerable  
231 change in the flow velocities, even at the upper boundary. Also, as Fig. 2 illustrates, the conditions above an obstacle  
232 might be slightly wetter than below an obstacle, but the variations in the pressure head and the water content fields is  
233 very small. We note that this general finding was equally applicable for all other pressure heads steps in the virtual  
234 MSUG experiment.

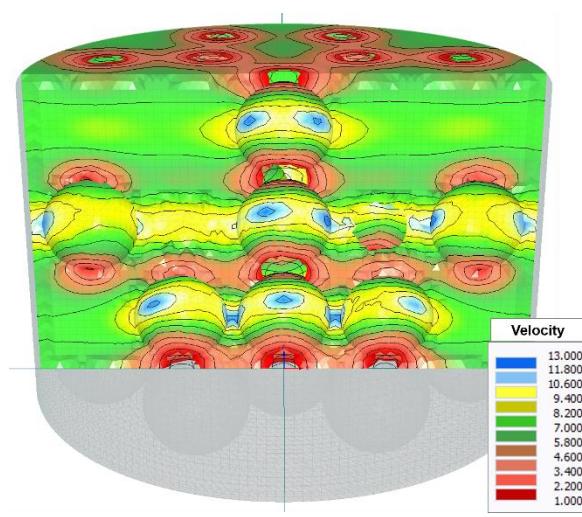
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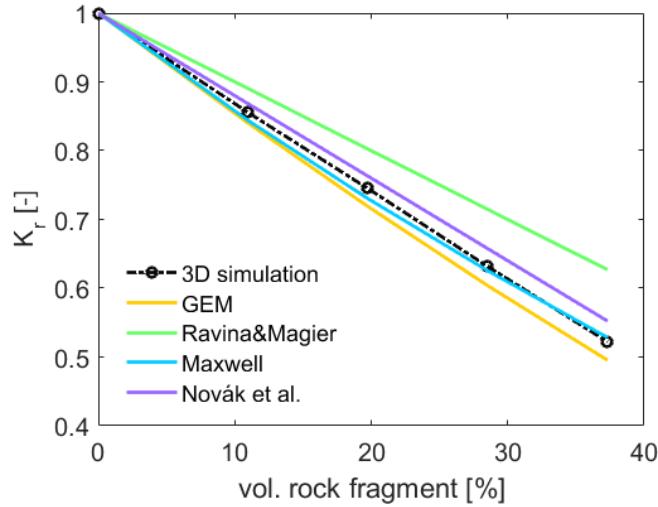
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238      *Figure 2: Visualization of the pressure head (cm), water content ( $\text{cm}^3 \text{ cm}^{-3}$ ), and velocity ( $\text{cm d}^{-1}$ ) in a 2D profile in the center of the soil column during the forward simulation of MSUG experiment.*  
239

240 **3.2. Comparison of the relative  $K_s$  of the scaling models and the 3D simulations under saturated conditions**

241 The dependency of the relative saturated hydraulic conductivity ( $K_r$ ) on the percentage of RF, calculated by different  
 242 scaling models and the obtained values from the first pressure in the virtual MSUG experiment is presented in Fig. 3.  
 243 The results of the models are shown up to the  $f = 37.3\%$ , which was the highest value of  $f$  simulated in 3D. However,  
 244 some of the evaluated models are theoretically valid for higher or lower values of  $f$ , e.g. 40 % for the Novák et al.  
 245 (2011) model and higher values for the GEM model (Naseri et al., 2020).



246

247 *Figure 3: Comparison of the values of  $K_r$  (-) from the virtual MSUG experiment in 3D (circles), and calculated by different scaling  
 248 models (solid lines) for  $f$  values up to 37.3 %. The dashed line connects the simulated data points of  $K_r$  shown by circles.*

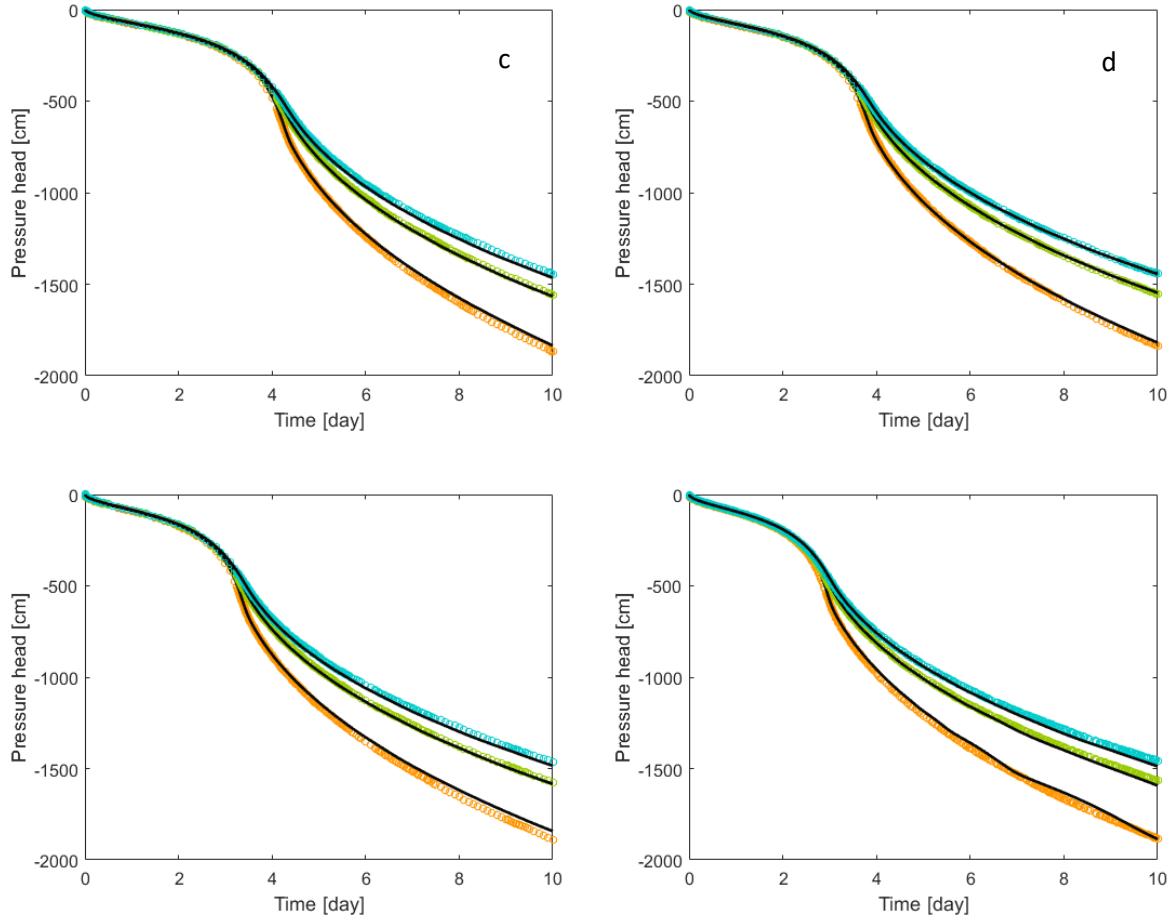
249 Obviously, the results of our simulations confirm a linear reduction of  $K_r$  with increasing the volume of RF up to  $f =$   
 250 37.3 % (v/v) in the soil. The numerically obtained values of  $K_r$  are shown by circles and connected by the dashed line  
 251 in Fig. 3. The dashed line has a slope of -1.29 representing a higher reduction rate of  $K_r$  compared to the scaling of  $K_r$   
 252 that would be proportional to  $f$ , expressed by Eq. (6) and predicted by the model of Ravina and Magier (1984) (solid  
 253 green line). This result supports the fact that even in a stony soil with spherical impermeable RF, the reduction in the  
 254 hydraulic conductivity is higher than the reduction of the average cross-sectional area (which is statistically equivalent  
 255 to the value of  $f$ ). Hlaváčiková et al. (2016) found an even higher value of -1.45 for spherical RF with a diameter of  
 256 10 cm. The model of Novák et al. (2011) performs better but also leads to a slight under-prediction of the reduction  
 257 of the effective saturated conductivity. The performance of this model could be improved by adjusting the parameter  $\alpha$   
 258 to match data of the 3D simulation, but doing this would lead to an unfair comparison with the other models. The two

259 models predicting a nonlinear relationship between the  $K_r$  and  $f$ , GEM and Maxwell, show similar results at low  
260 contents of RF up to 10 %, with minor differences in outputs of the models. Among all of the evaluated models of  
261 scaling  $K_s$ , the Maxwell model yields the closest match to the numerically identified values of  $K_r$ .

262 We note that these results may differ in natural soils, where an increase of the saturated hydraulic conductivity might  
263 be expected because of macropore flow in lacunar pores at the interface between background soil and RF (Beckers et  
264 al., 2016; Hlaváčiková et al., 2019, Arias et al., 2019). We have not included such a process in our 3D simulations.

265 **3.3. Inverse modeling results for effective hydraulic properties**

266 The observed and fitted time series of the pressure heads at the three representative observation points is shown in  
267 Fig. 4 for the simulated experiments of the four cases with different values of  $f$ . In each case, the fitted pressure heads  
268 at the three depths of the column (2.5, 5.0 and 7.5 cm) match well with the time series of the corresponding data from  
269 the 3D virtual experiments. Specifically, the match of the pressure heads at 7.5 and 5.0 cm is excellent, whereas there  
270 are slight systematic deviations at the uppermost level at the later stage of the virtual EVA experiment.



271

272

273 *Figure 4: The time series of the 3D-simulated (circles) and 1D fitted (solid lines) pressure heads at the observation points in three*  
 274 *depths of the stony soil columns with the  $f$  values a) 11.0 %, b) 19.8 %, c) 28.5 %, and d) 37.3 %. The observation depths are*  
 275 *indicated by different color codes of orange (upper, 2.5 cm), green (middle, 5.0 cm) and blue (lower, 7.5 cm from top).*

276

277 Table 1 shows the values of the root mean square error (RMSE) and mean absolute error (MAE) between the observed  
 278 and fitted time series of the pressure heads at three observations points for values of  $f$ . According to the Table 1, the  
 279 fit is best for the lower  $f$  and in the middle of the column. The highest deviations occur for the highest  $f$  but there is  
 280 no clear trend. Overall, the values of RMSE and MAE are in an acceptable range regarding the observed values of  
 281 pressure heads up to -2000 cm. This indicates that the time series of the pressure heads at multiple depths generated  
 282 by the 3D simulations of EVA experiments can be described successfully by the 1D Richards equation assuming a  
 283 homogenous system with effective SHP.

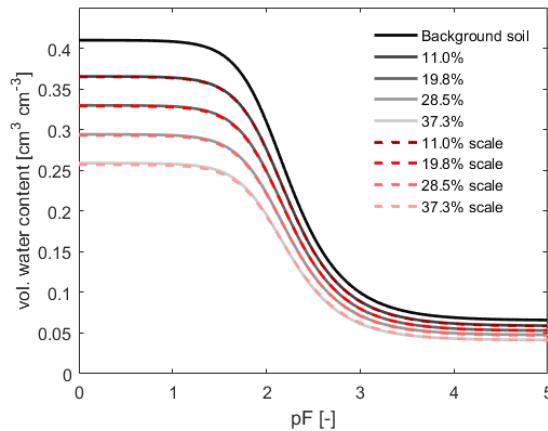
284

285 *Table 1: The values of RMSE and MAE between the observed and fitted pressure heads, in three observation points for different*  
 286 *values of  $f$ .*

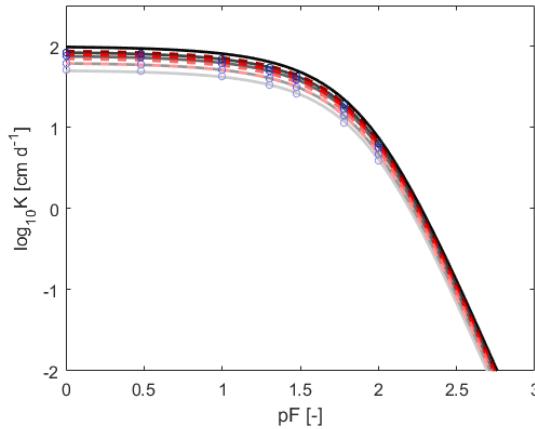
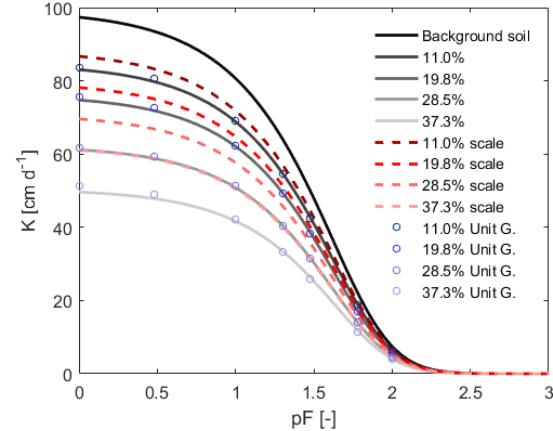
Criteria	Observation point	Volumetric fraction of RF, $f$ (%)		
		11.0	19.8	28.5
RMSE	upper	12.8	6.5	12.6
	middle	4.9	2.8	5.3
	lower	9.1	3.1	8.7
MAE	upper	10.0	4.9	9.4
	middle	4.0	2.1	4.5
	lower	7.0	2.6	6.8
				11.0

287

288 The identified SHP are presented in Fig. 5. The solid lines in the Figure show the WRC and HCC of the virtual stony  
 289 soils obtained by inverse simulation (except the solid black lines, which are the WRC and HCC of the background  
 290 soil), the dashed lines represent the scaled WRC by Eq. (5) and HCC by Eq. (6), and the circles on the HCC plots  
 291 represent the discrete data points of hydraulic conductivity obtained by the virtual MSUG experiment. The WRC and  
 292 HCC are presented on a  $pF$  scale, which is defined as  $pF = \log 10 (|h|)$ , in which  $h$  is the pressure head in cm  
 293 (Schofield, 1935). The van-Genuchten model parameters of the background soil and stony soils are shown in Table 2.



294



295

296 *Figure 5: The WRC and HCC of the background soil (solid black line), and identified effective WRC (left) and HCC (right) of the*  
 297 *virtual stony soils (solid gray lines) with different values of  $f$ . The HCC are presented also in the logarithmic scale. The dashed*  
 298 *lines show the effective WRC and HCC calculated by the models of Bouwer and Rice (1984) and Ravina and Magier (1984). The*  
 299 *circles on the HCC present the data points of hydraulic conductivity obtained by the virtual MSUG experiment in near-saturated*  
 300 *conditions up to the  $pF \approx 2$ .*

301

302 *Table 2: The van-Genuchten model parameters of the SHP of the background soil and of the inversely determined effective SHP of*  
 303 *the virtual stony soils with different values of  $f$ .*

Parameter	Unit	Volumetric fraction of RF, $f$ (%)				
		Background soil	11.0	19.8	28.5	37.3
$\theta_s$	(cm³ cm⁻³)	0.410	0.365	0.330	0.294	0.259
$\theta_r$	(cm³ cm⁻³)	0.065	0.059	0.053	0.048	0.041
$\alpha$	(cm⁻¹)	0.010	0.010	0.010	0.010	0.010
$n$	(-)	2.000	2.007	2.011	2.014	2.037
$K_s$	(cm d⁻¹)	100.0	84.7	76.2	62.3	50.4
$\tau$	(-)	0.50	0.43	0.47	0.42	0.38

304

305 According to the Fig. 5 and Table 2 the value of the shape parameter  $\alpha$  is independent of the value of  $f$  and the change  
306 in the value of  $n$  is negligible (but might be systematic). The inversely identified WRC and the predictions from the  
307 Bouwer and Rice (1984) scaling model match almost perfectly for all the RF contents. In agreement with this, there  
308 is also an excellent agreement of the values of the saturated ( $\theta_s$ ) and residual water contents ( $\theta_r$ ) ( $\text{cm}^3 \text{cm}^{-3}$ ) between  
309 scaled and identified WRC. The values of  $\theta_s$  and  $\theta_r$  in the WRC are directly related to the values of  $f$  and the respective  
310 values of the background soil and the WRC is scaled by this factor over the whole range of soil water pressure head.

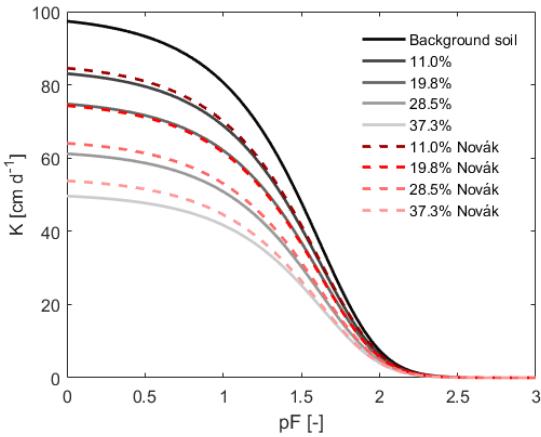
311 Similar to the WRC, an increase of  $f$  reduces the hydraulic conductivity over the whole range of pressure head covered  
312 by the virtual experiments. However, in contrast to the WRC, the simple scaling model based on Eq. (6) cannot  
313 describe the reduction in HCC. Figure 5 shows that the model of Ravina and Magier (1984, dashed lines)  
314 underestimates the reduction of the effective HCC for all rock contents. The reason might be related to the local  
315 variations of the flow velocity in the soil column. It was shown in Fig. 2 that the variations in the water flow velocity  
316 might be considerable. The nonlinearities in the flow field and changes in the local conductivities, together with an  
317 increased average flow path length, force a stronger overall conductivity reduction. The arrangement of RF thus might  
318 affect the reduction in hydraulic conductivities, leading to different conductivities at the same value of  $f$  (Naseri et  
319 al., 2020). The degree depends on how the flow area is altered in the soil column due to the presence of RF (Fig. 1  
320 and 2). This result is in agreement with Novák et al. (2011) who reported a higher reduction in conductivity compared  
321 to a reduction that is proportional to the rock fragments content. Furthermore, it may also differ alter depending on  
322 the characteristics of RF such as their size, shape and orientation towards flow (Novák et al., 2011).

323 We had to include the data points of hydraulic conductivity from the virtual MSUG experiment in the inverse objective  
324 function to get a precise identification of the HCC obtained by inverse modeling near saturation. The information  
325 content from the virtual EVA experiment gives a unique identification only when the flux rate in the system reaches  
326 the magnitude of the unsaturated hydraulic conductivity, which is for many soils around  $pF = 1.5$  to  $pF = 2$  (Peters  
327 and Durner, 2008). Although there are some discrepancies visible near saturation for the case with a high value of  $f$ ,  
328 the resulting values of hydraulic conductivity from the virtual MSUG experiment and the inversely identified HCC  
329 using the virtual EVA experiment join well around  $pF = 2$  for all of the values of  $f$ . Therefore, the HCC could be  
330 described successfully from the saturation up to  $pF = 3$  using the inverse modeling of the virtual EVA experiment  
331 with added  $K$  support points from the virtual MSUG experiment. The overall results suggest that the effective

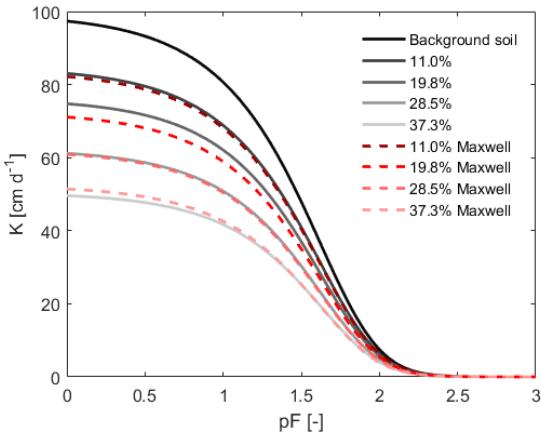
332 hydraulic parameters of stony soils could be obtained by the corresponding real experiments and the result is robust  
333 for both, WRC an HCC, even if the uncertainty in the identified HCC is higher than that of the WRC (Singh et al.,  
334 2020; 2021).

335 **3.4. Evaluation of the Novák, Maxwell and GEM models using the identified HCC**

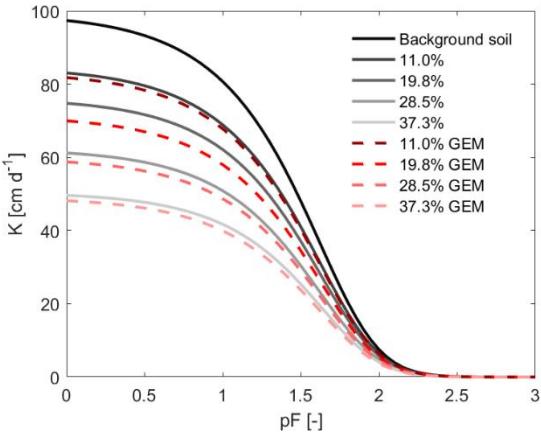
336 As stated above, the model of Ravina and Magier (1984) which is a linear scaling approach of the hydraulic  
337 conductivity (Eq. 6) underestimates the reduction of conductivity in the stony soil. We used the identified HCC as a  
338 benchmark to evaluate and compare more advanced models of scaling HCC, namely the Novák, Maxwell and GEM  
339 models (Eq. 7, 8 and 9). Figure. 6 illustrates the calculated HCC of stony soils with different values of  $f$  using these  
340 models of scaling HCC and compares them to the identified HCC by the inverse modeling.



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Figure 6: Evaluation of the Novák, Maxwell and GEM models of scaling HCC of stony soils using the identified HCC as a benchmark. The HCC in each case were obtained for  $f=11.0, 19.8, 28.5$  and  $37.3\% (v/v)$ . The inverse identified curves are shown in solid lines and the model results in dashed lines. The value of  $f_c$  in the GEM model was set as 0.982 with the corresponding shape parameter  $t=1.473$  and the parameter  $\alpha = 1.2$  was selected for the Novák model.

348

349 The calculated HCC by the three models are in general in good agreement with the identified HCC in the observed  
350 range of pressure heads. All three models result in a more realistic estimate of the HCC compared to the simple linear  
351 scaling approach. While the model of Novák slightly underestimates the identified HCC for all the four RF contents,  
352 the results are contrary for the GEM model where the reduction in the hydraulic conductivity is overestimated. The  
353 Maxwell model shows the same results as GEM model except for the stony soil with  $f = 37.3\%$  where it  
354 underestimates the HCC.

355 In order to compare the performance of the three models, the average deviation ( $d_{avg}$ ) between the calculated and  
356 identified HCC (logarithmic scale) was calculated to quantify the error of each model in the  $pF$  range 0 to 3 (Table  
357 3). The signs of numbers in Table 3 represent the tendency of the model in over- or underestimating the identified  
358 hydraulic conductivities. The negative sign means the model underestimates the reduction of hydraulic conductivity.

359 *Table 3: Performance of the Novák, Maxwell and GEM models quantified by the average deviation of  $\log_{10}(K)$  ( $d_{avg}$ ) for  
360 different values of  $f$ .*

model	Volumetric fraction of RF, $f$ (%)			
	11.0	19.8	28.5	37.3
Novák	-0.0068	-0.0028	-0.0179	-0.0231
Maxwell	0.0056	0.0162	0.0039	-0.0038
GEM	0.0077	0.0234	0.0197	0.0248

361  
362 Table 3 confirms the qualitative tendency of underestimation of the conductivity reduction by the Novák model and  
363 the overestimation by the GEM and Maxwell models, but also shows that the difference between the three models is  
364 not large and probably not of relevance in practice (the GEM model at the high value of  $f$ , which has the highest  
365 deviation, corresponds to a relative mismatch of  $K$  of 6 %). However, despite the potential of the three models in  
366 predicting the HCC of stony soils, we think they require further evaluations using field measured data of hydraulic  
367 conductivity in different experimental conditions.

368 **4. Conclusions**

369 In this study, we created virtual stony soils with different volumetric fractions of RF in 3D and identified their effective  
370 SHP from saturation up to  $pF = 3$  by inverse modeling of virtual EVA and MSUG experiments in 1D. We used the  
371 identified SHP to investigate the performance of the available scaling models in stony soils, namely the Bouwer and

372 Rice (1984) model of scaling the WRC, and Ravina and Magier (1984), Maxwell (Peck and Watson, 1979), Novák  
373 (Novák et al., 2011), and GEM (Naseri et al., 2020) models for scaling HCC of stony soils.

374 Our results show that the boundary fluxes and the internal system states in the virtual 3D EVA experiments,  
375 represented by the observed time series of pressure heads at multiple depths, could be matched well by 1D simulations,  
376 and the effective WRC and HCC of the considered stony soils were determined accurately. Comparison with the  
377 scaling models showed that by assuming a homogeneous background soil and impermeable RF, the effective WRC  
378 can be calculated from the WRC of the background soil using a simple correction factor equal to the volume fraction  
379 of background soil,  $(1 - f)$ . That is a result with practical implications in obtaining WRC of stony soils. In addition,  
380 the scaling results for HCC were promising. Our results confirmed that the reduction in  $K_r$  was stronger than calculated  
381 by a simple proportionality to  $(1 - f)$ . The three models of Novák, Maxwell, and GEM consider that and performed  
382 adequately in predicting the effective HCC of stony soils. The Maxwell model matched the numerical results best.

383 Care must be taken before generalizing these results to arbitrary conditions, e.g., highly dynamic boundary conditions  
384 with sequences of precipitation and higher and lower evaporation rates, which might yield different results due to the  
385 occurrence of non-equilibrium water dynamics and hysteresis. For real stony soils, changes in the pore size distribution  
386 of the background soil may result from the presence of RF (Sekucia et al., 2020) with corresponding consequences  
387 for effective SHP. This influence was reported to be more common in compactable soils with a shrinkage-swelling  
388 potential (Fiès et al., 2002). In highly stony soils, where RF are not embedded completely in the background soil, the  
389 existence of effective SHP is still an open question. Finally, the impact of arrangement and size of RF on evaporation  
390 dynamics and effective SHP needs to be understood. Tackling these problems requires a combination of experimental  
391 and modelling approaches.

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