Test of the Rosetta Pedotransfer Function for Saturated Hydraulic Conductivity

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ABSTRACT

Simulation models are tools that can be used to explore, for example, effects of cultural practices on soil erosion and irrigation on crop yield. However, often these models require many soil related input data of which the saturated hydraulic conductivity (K_s) is one of the most important ones. These data are usually not available and experimental determination is both expensive and time consuming. Therefore, pedotransfer functions are often used, which make use of simple and often readily available soil information to calculate required input values for models, such as soil hydraulic values. Our objective was to test the Rosetta pedotransfer function to calculate K_s . Research was conducted in a 64-ha field near Lamesa, Texas, USA. Field measurements of soil texture and bulk density, and laboratory measurements of soil water retention at field capacity (-33 kPa) and permanent wilting point (-1500 kPa), were taken to implement Rosetta. Calculated values of K_s were then compared to measured K_s on undisturbed soil samples. Results showed that Rosetta could be used to obtain values of K_s for a field with different textures. The Root Mean Square Difference (RMSD) of K_s at 0.15 m soil depth was 7.81 × 10⁻⁷ m·s⁻¹. Further, for a given soil texture the variability, from 2.30 × 10⁻⁷ to 2.66 × 10⁻⁶ m·s⁻¹, of measured K_s was larger than the corresponding RMSD. We conclude that Rosetta is a tool that can be used to calculate K_s in the absence of measured values, for this particular soil. Level H5 of Rosetta yielded the best results when using the measured input data and thus calculated values of K_s can be used as input in simulation models.

Keywords: Rosetta; Saturated Hydraulic Conductivity; K_s; Pedotransfer Function

1. Introduction

In order to model soil physical processes related to soilwater content, it is important to know the hydraulic properties of the soil [1]. Vigiak *et al.* [2] also remarked the importance of characterizing soil hydraulic parameters in order to understand the occurrence and movement of overland flow at field, hill-slope, and catchment scale. These soil hydraulic properties include the saturated (K_s) and unsaturated hydraulic conductivity, and the water retention curve. Furthermore, in any modeling study on water flow and solute transport in soils, the water retention and hydraulic conductivity functions for soil horizons in the profile are crucial input parameters [3]. Examples of two simulation models that require soil hydraulic properties as inputs are the Energy Water Balance Model [4,5] and HYDRUS [6,7]. In addition, soil hydraulic properties are a required input in models used to calculate water runoff and soil erosion, e.g. Precision Agricultural Landscape Modeling System [8-10] and examples of other models are given by Aksoy and Kavvas [11].

Although the soil hydraulic properties can be measured directly, this practice is both costly and time-consuming, and sometimes results obtained are unreliable because of the associated soil heterogeneity and experimental errors [12,13]. When large areas of land are under study, it is virtually impossible to perform enough measurements to be meaningful, indicating the need for an inexpensive and rapid way to determine soil hydraulic properties [14]. For example, some research results indicated that in a 10-ha field, 1300 measurements would have to be made in order to accurately measure saturated hydraulic conductivity to within 10% of the mean value [13]. Furthermore, Stroosnijder [15] indicated that measurements alone would provide data difficult to extrapolate in time and space, meaning that erosion in large fields would not



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be known without prediction technologies, but also that predictions need measurements to work. Therefore, many indirect methods such as pedotransfer functions (PTFs) have been developed to reduce the effort and cost [16]. These PTFs are predictive functions of certain soil properties estimated from other simpler measured soil proper-ties [17]. They can be used as inputs to models in order to reduce costs and accelerate the investigations. A detailed review of PTFs is given by Wösten *et al.* [3] and scaling of soil physical properties in relation to models is given by Pachepsky *et al.* [18]. In summary, many efforts have been made to develop PTFs at large scale, but little has been done to evaluate the performance of Artificial Neural Networks (ANNs) in estimating saturated K_s at a landscape scale [19].

A soil hydraulic property that is often a required input to simulation models is the saturated hydraulic conductivity, K_{s} . It is one of the most important soil physical properties for determining infiltration rate and other hydrological processes [20]. In general, the K_s refers to the capacity of the soil to drain water and gives information about the presence of disruptive soil strata, and the correlation between the permeability and other soil characteristics. The geometry of the complex pores that depend on texture, structure, viscosity and density, determine the K_s . In hydrologic models, this is a sensitive input parameter and is one of the most problematic measurements at field-scale in regard to variability and uncertainty [21]. The K_s is known to be one of the most variable of all soil physical properties, varying up to 10 orders of magnitude for different geo-materials [22]. Thus, the objective of this study was to determine the applicability of the Rosetta pedotransfer function [14] to calculate K_s for different soil textures of a large field in the Texas Southern High Plains. The calculated values of K_s were evaluated by comparing them to laboratory-measured values. Furthermore, the differences between measured and calculated values of K_s were evaluated using two statistical parameters, the mean square deviation [23] and the Nash Sutcliffe Efficiency parameter [25]. A use of the Rosetta PTF is to assist researchers in studying the hydrological processes that could lead to better water management practices in the region of study. This is important because in the Texas Southern High Plains the combination of common droughts and a declining water table from the Ogallala Aquifer are a challenge to manage the irrigation of crops and establish crops under dryland conditions [5,25].

2. Materials and Methods

2.1. Study Area

The study area is located in the region known as Llano Estacado, near Lamesa, which is a small town, popula-

tion of close to 10,000, located in Dawson County, Texas, USA (**Figure 1**). The coordinates are $32^{\circ}32'7.4''N$; 101°46′8.24″W; and, an elevation of 833 m above sea level with a semi-arid climate. The field is 63.8 ha and has an average slope of 0.33%, although parts of the field have slopes in excess of 5%. In this region, about half the cultivated land is irrigated from an underground aquifer, called the Ogallala, which is classified, as non-rechargeable [25]. The predominant soil series at this site is an Amarillo fine sandy loam, a fine-loamy, mixed, superactive, thermic, Aridic Paleustalf [26]. These soils are characterized by a high content of Ca and Mg carbonates (pH > 7), low organic matter (<2 g·kg⁻¹), and are classified as moderately permeable.

2.2. Soil Sampling

To determine the K_s as well as the water retention curve, 18 undisturbed soil samples were collected using a gouge auger, from 0.10 - 0.15 m depth from the surface at the locations shown in **Figure 2**. Sampling was done on 15 July 2009. The sampler rings were 0.05 m in diameter by 0.051 m long. For the purpose of this study, only one sample was taken at each of the 18 locations; therefore, they can be considered as replicates.

To determine the textural classes, 36 soil cores were sampled over the field on the 15 July 2009 at the locations shown in Figure 3, which were geo-referenced using a Global Positioning System (Model 4700 Dual Channel RTK system, Trimble, Sunnyvale, CA). These samples were taken with a commercial tractor-mounted hydraulic core sampler system (Giddings Machine Company, Model HDGSRTS, Windsor, CO). This system pushed into the soil a tube-sampler with a plastic sleeve inside, i.e., 1.2 m long and 0.05 m in diameter. Once the sampler was removed from the soil the sleeve inside the core was extracted and the undisturbed soil sample obtained. Thereafter, the soil cores were stored in a coldroom at a temperature of 5°C. These sampling locations were selected using the NRCS (2008) soil map [26] as a guide to ensure that all the different soil types within the field were included in the sampling scheme. The soil cores were used to identify the lower and upper depth of soil horizons (Table 1). This soil was then oven-dried at 105 for 24 hrs and the dry samples were ground with a mill°C and sieved to 2 mm.

2.3. Rosetta Description

Rosetta is an algorithm that calculates soil water retention parameters, K_s and unsaturated hydraulic conducti-

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Figure 1. Location of the field-study, Lamesa, Dawson county, Texas, USA. (Source: Self composition on a Google-USDA farm service agency image).



Figure 2. Sampling locations in the field where 18 soil cores were obtained and saturated hydraulic conductivity (K_s) , and water retention were measured. (Source: Self composition on a Google-USDA farm service agency image).



Figure 3. Spatial distribution of the textural classes in Lamesa, Texas.

 Table 1. Upper and lower depth limits of the soil horizons

 for the 18 soil cores taken at the field in Lamesa, Texas.

Soil horizon	Upper limit (m)	Lower limit (m)	
1Ap	0.0	0.3	
2Ap	0.1	0.3	
1Bt	0.2	0.5	
2Bt	0.3	1.2	
3Bt	0.4	1.2	
Btk	0.5	1.2	
Btkk	0.5	1.2	

vity using hierarchical PTFs based on five levels of input data [14]. It is of great practical use, allowing flexibility for the user towards the required input data [27]. The first level (H1) consists of a lookup table that provides average parameters for each of the USDA textural classes. The second level (H2) uses values from H1 plus sand, silt, and clay fractions as inputs, and provides a hydraulic parameter that varies continuously with texture. The third level (H3) includes the predictors used in level H2 and the soil dry-bulk density (ρ_d). The fourth level (H4) uses H3 and soil volumetric water content (θ) at a water suction of -33 kPa. The last level (H5) consists of all the other parameters, H4, plus the θ at a water suction of -1500 kPa. While H1 is a simple table with average hvdraulic parameters for each textural class, all other models involve a combination of neural networks and the bootstrap method [14].

In Rosetta, the relation between θ and water suction (h), *i.e.* water retention $[\theta(h)]$, as well as the saturated and unsaturated hydraulic conductivity, are described with the well known Mualem-Van Genuchten equation [28,29] and is given by:

$$\theta(h) = \theta_r + \frac{(\theta_s - \theta_r)}{\left[1 + (\alpha \times h)^n\right]^m}$$
(1)

where $\theta(h)$ is the soil volumetric water content (m³·m⁻³) at suction h (cm); θ_s and θ_r are the saturated and the residual water content (m³·m⁻³) at h = 0 cm and -15,000 cm, respectively; α (> 0 in cm⁻¹) is related to the inverse of the air entry suction; and n (> 1) is a measure of the pore-size distribution and m = 1 - 1/n. The unsaturated hydraulic conductivity, $K(S_e)$, is described with the Mualem-Van Genuchten model as:

$$K(S_e) = K_0 S_e^L \left[1 - \left(1 - S_e^{\left(\frac{n}{n-1}\right)} \right)^m \right]^2$$
(2)

where K_0 is a fitted value of K at saturation (cm·d⁻¹), which is similar but not considered equal to K_s , and L is a pore connectivity factor (negative in most cases). The effective saturation (S_e) is given by:

$$S_{e} = \frac{\left(\theta(h) - \theta_{r}\right)}{\left(\theta_{s} - \theta_{r}\right)} = \left[\frac{1}{1 + \left(\alpha \times h\right)^{n}}\right]^{m}$$
(3)

Therefore, the relative hydraulic conductivity $K_r(h)$ is given by:

$$K_{r}(h) = \frac{\left[1 - \left(\alpha \times h\right)^{n-1} \left[1 + \left(\alpha \times h\right)^{n}\right]^{m}\right]^{2}}{\left[1 + \left(\alpha \times h\right)^{n}\right]^{\frac{m}{2}}}$$
(4)

which is a function given the quotient of hydraulic con-

ductivity function, K(h) to saturated hydraulic conductivity, K_s [27]. In summary, the seven parameters calculated with the Rosetta PTF (Equations (1)-(4)) are: θ_r , θ_s , α , n, K_s , K_0 , and L.

2.4. Model Performance Evaluation

Calculated values of K_s obtained with Rosetta were compared to corresponding measured values and evaluated using two statistical parameters. First, by using the Root Mean Squared Difference (RMSD) and second, by using the Nash Sutcliffe Efficiency (NSE). The RMSD, gives the mean difference between measured and calculated values of K_s and is calculated as [23]:

RMSD =
$$\sqrt{\frac{1}{n} \sum_{i=1}^{n} (x_i - y_i)^2}$$
 (5)

where x_i is the measured value of K_s and y_i is the corresponding calculated value of K_s obtained with Rosetta.

The NSE compares measured and calculated values of K_s and is given by [24]:

NSE = 1 -
$$\left[\frac{\sum_{i=1}^{n} (Y_{i}^{m} - Y_{i}^{c})^{2}}{\sum_{i=1}^{n} (Y_{i}^{m} - Y_{i}^{mean})^{2}}\right]$$
 (6)

where Y_i^m is measured value of K_s and Y_i^c is the corresponding calculated value of K_s , Y_i^{mean} is the average of measured values of K_s , and n is number of measurements. Values of NSE may range from $-\infty$ to 1. A value of NSE = 1, corresponds to a perfect match of calculated compared to the measured values of K_s . Conversely, an efficiency of 0 (NSE = 0) indicates that the Rosetta calculations of K_s are as accurate as the mean of the measured data; whereas, an efficiency <0 (NSE < 0) occurs when the measured mean is a better predictor than the model or, in other words, when the residual variance, described by the nominator in Equation (6), is larger than the data variance, described by the denominator [24].

2.5. Measurements

Textural Analysis. Clay, silt and sand fraction of all 36-soil samples were determined using the hydrometer method [30]. This method uses the Navier-Stokes equation to calculate soil particles in suspension in an infinite soil column and is adequate for textural class identification, but cannot be used to accurately define the particle size [31]. Nevertheless, it provides a reasonable input to Rosetta [14].

Saturated Hydraulic Conductivity (K_s). The K_s was measured under laboratory conditions using the constant and falling head methods [31] on undisturbed soil samples. For both methods, K_s were measured using a laboratory permeameter [32]. In some soil samples, water

flowed in the range of operation for the falling head method, while in other samples the water flowed in the range of the constant-head method. For this reason, both methods were used and thereafter matched to the K_s of each soil sample. Our first approach was to use the constant head method and if K_s was $< 1.16 \times 10^{-7}$ m s⁻¹, the falling head method was selected. As previously described, the undisturbed soil samples used for the measurement of K_s were collected using a gouge auger at the locations shown in Figure 2. To avoid the possible disintegration of soil particles in each sample, which would cause silting of the pores and a reduction of the flow and K_s , we used ultrahigh-pure water that was autoclaved at 120°C for 30 minutes. The solution was saturated with calcium sulphate (CaSO₄·2H₂O) and toluene was added to eliminate microorganisms.

Water Retention Determination. The relation between soil volumetric water content (θ) and water suction (h) at saturation (h = 0) and field capacity (h = -33 kPa) was measured using a sand/kaolin box (pF-Determination, Eijkelkamp, Giesbeek, The Netherlands), that operates in the 0 - 50 kPa range [33]. For these two measurements we used the undisturbed soil samples that were also used to measure K_s . Weights of the soil samples, before and after drying, and ring volumes were used to calculate the soil gravimetric water content (θ_g , kg·kg⁻¹) and the corresponding value of the soil dry-bulk density (ρ_d , kg·m⁻³) was used to convert θ_g to θ , assuming a density of water equal to 1000 kg·m⁻³.

For the soil water content at wilting point, pF = 4.18 ($h = -1500 \text{ kPa} \approx -15 \text{ bar}$), the water in the soil is retained in small pores, and thus the retention of water is dominantly influenced by texture. For this reason, disturbed soil samples can be used for this determination, which were saturated with water for 2 - 3 days and placed on a plate (Pressure-Membrane Apparatus, Eijkelkamp, Giesbeek, The Netherlands) and a pressure of 1500 kPa was applied for 3 days to the three replicates [34]. Afterwards soil samples were weighed (wet mass), oven-dried at 105°C and weighed again (dry mass). With these values the θ_g water content, was calculated for a pF = 4.18, and converted to θ using the ρ_d of the undisturbed samples used in the sand/kaolin box.

3. Results and Discussion

3.1. Measurements

There were three textural classes corresponding to the 18 spots where undisturbed soil samples were taken (**Figure 2**) and these were: clay loam, sandy clay loam, and sandy clay. Of the total number of soil samples, there were thirteen samples of sandy clay loam, four of clay loam and one sandy clay. The ESRI software (version 9.2) of ArcGIS (2006) was used as the Geographic Information

System (GIS) processing software, and the inverse distance weighed interpolation was used to extend the measured values of texture across the entire field. **Figure 3** shows the textural classes at a 0.15 m depth, showing that most of the field is classified as a sandy clay loam.

The field has five distinct areas (see **Figure 2**) with oil wells and access roads that are not cultivated. These areas were included in our analysis as they contribute to the hydrological processes of the entire field, and were assigned a textural class of sandy clay.

Values of soil volumetric water content (θ) as a function of h measured with the kaolin/sand box and pressure membrane are shown in **Table 2** using the 18 undisturbed soil samples taken from the 0.10 - 0.15 m soil layer. These results show that the sandy clay retains more water at saturation and this is probably related to the porosity, which is also the highest (45.5%) among the soil samples taken. The soil θ at h = -33 kPa and h = -1500 kPa given in **Table 2** were used as inputs to the Rosetta PTF.

Values of soil dry-bulk density (ρ_b) for each textural class and its porosity, calculated assuming a particle density of 2650 kg·m⁻³ are shown in **Table 3**. The sandy clay had the lowest ρ_b and thus the highest porosity. High values of ρ_b in the sandy clay loam samples lowered the calculated porosity affecting the hydraulic properties of the corresponding horizon.

Measured values of saturated hydraulic conductivity (K_s) for each of the 18 locations of **Figure 2** are given in **Table 4** and plotted in **Figure 4**. Values of K_s ranged

Table 2. Average soil volumetric water content (θ , m³·m⁻³) and standard deviation (SD) as a function of h at 0, -33 and -1500 kPa for the three textural classes in Lamesa, Texas.

Textural class (number of samples)	Soil volumetric water content $(\theta, m^3 \cdot m^{-3})$			
	0	-33	-1500	
	<i>h</i> (kPa)	<i>h</i> (kPa)	<i>h</i> (kPa)	
Clay loam (Cl) (4)	0.45	0.30	0.17	
	(0.03)	(0.02)	(0.05)	
Sandy clay (SC) (1)	0.49	0.27	0.14	
	(0.00)	(0.00)	(0.00)	
Sandy clay loam (SCL) (13)	0.45	0.29	0.15	
	(0.04)	(0.02)	(0.05)	

Table 3. Average measured soil dry-bulk density (ρ_b , kg·m⁻³) and calculated porosity (%), and standard deviation (SD) for the textural classes in Lamesa, Texas.

Textural class (samples)	$ ho_{ m b} \ (m kg\cdot m^{-3})$	Porosity (%)
Clay loam (4)	1620 (0.05)	39.0 (2.00)
Sandy clay (1)	1440 (0.00)	45.5 (0.00)
Sandy clay loam (13)	1640 (0.09)	38.1 (3.50)

Table 4. Measured and calculated values of K_s obtained using the five hierarchical (H) levels of the Rosetta PTF. In the texture (USDA) column S is for sand, C is for clay and L is for loam, and all values of K_s have units of m s⁻¹.

Location	Texture (USDA)	<i>K</i> _s measured	<i>K</i> _s -H1 USDA textural class	<i>K</i> _s -H2 texture	K_s -H3 texture + ρ_b	K_s -H4 texture + ρ_b + θ_{33}	K_s -H5 texture + ρ_b + θ_{33} + θ_{1500}
1	SCL	3.50×10^{-7}	$1.5 imes 10^{-6}$	$1.5 imes 10^{-6}$	9.2×10^{-7}	5.7×10^{-7}	6.4×10^{-7}
2	CL	1.30×10^{-6}	9.5×10^{-7}	$7.9 imes 10^{-7}$	6.9×10^{-7}	$8.3 imes 10^{-7}$	$9.0 imes 10^{-7}$
3	SCL	3.50×10^{-7}	$1.5 imes 10^{-6}$	$1.2 imes 10^{-6}$	6.7×10^{-7}	5.2×10^{-7}	$5.3 imes 10^{-7}$
4	SCL	$3.50\times 10^{\text{-8}}$	$1.5 imes 10^{-6}$	$1.1 imes 10^{-6}$	4.0×10^{-7}	$2.9 imes 10^{-7}$	3.7×10^{-7}
5	SCL	2.30×10^{-6}	$1.5 imes 10^{-6}$	$8.3 imes 10^{-7}$	9.4×10^{-7}	$2.2 imes 10^{-6}$	2.1×10^{-6}
6	SCL	$1.40 imes 10^{-6}$	$1.5 imes 10^{-6}$	9.6×10^{-7}	$1.3 imes 10^{-6}$	$2.4 imes 10^{-6}$	$2.3 imes 10^{-6}$
7	CL	$2.30 imes 10^{-7}$	$9.5 imes 10^{-7}$	$1.8 imes 10^{-6}$	4.4×10^{-7}	3.4×10^{-7}	$3.6 imes 10^{-7}$
8	SC	3.70×10^{-6}	$1.3 imes 10^{-6}$	$8.8 imes 10^{-7}$	1.6×10^{-6}	4.9×10^{-6}	4.1×10^{-6}
9	CL	$2.30 imes 10^{-7}$	9.5×10^{-7}	$8.3 imes 10^{-7}$	7.0×10^{-7}	6.5×10^{-7}	7.1×10^{-7}
10	SCL	5.80×10^{-7}	$1.5 imes 10^{-6}$	$1.1 imes 10^{-6}$	9.6×10^{-7}	$1.0 imes 10^{-6}$	$1.0 imes 10^{-6}$
11	SCL	6.90×10^{-7}	$1.5 imes 10^{-6}$	$9.5 imes 10^{-7}$	6.1×10^{-7}	8.8×10^{-7}	8.9×10^{-7}
12	SCL	2.70×10^{-6}	$1.5 imes 10^{-6}$	$7.8 imes 10^{-7}$	$2.7 imes 10^{-7}$	$2.8 imes 10^{-7}$	$2.9 imes 10^{-7}$
13	CL	2.20×10^{-6}	9.5×10^{-7}	7.6×10^{-7}	$7.0 imes 10^{-7}$	1.9×10^{-6}	$1.8 imes 10^{-6}$
14	SCL	1.70×10^{-6}	$1.5 imes 10^{-6}$	$1.4 imes 10^{-6}$	$7.1 imes 10^{-7}$	$3.2 imes 10^{-6}$	$3.0 imes 10^{-6}$
15	SCL	1.30×10^{-6}	$1.5 imes 10^{-6}$	$1.1 imes 10^{-6}$	$1.8 imes 10^{-6}$	$2.5 imes 10^{-6}$	$2.4 imes 10^{-6}$
16	SCL	2.30×10^{-7}	$1.5 imes 10^{-6}$	$1.0 imes 10^{-6}$	4.9×10^{-7}	5.2×10^{-7}	$5.9 imes 10^{-7}$
17	SCL	$1.04 imes 10^{-6}$	$1.5 imes 10^{-6}$	1.1×10^{-6}	$8.3 imes 10^{-7}$	$1.6 imes 10^{-6}$	1.6×10^{-6}
18	SCL	1.30×10^{-6}	$1.5 imes 10^{-6}$	7.6×10^{-7}	$7.4 imes 10^{-7}$	$1.7 imes 10^{-6}$	1.6×10^{-6}



Figure 4. Measured values of saturated hydraulic conductivity (K_s , m·s⁻¹) for the 18 soil samples of the field in Lamesa, Texas.

from a low of $3.5 \times 10^{-8} \text{ m} \cdot \text{s}^{-1}$ at location 4, a sandy clay loam texture, to a high of $3.7 \times 10^{-6} \text{ m} \cdot \text{s}^{-1}$ at location 8, a sandy clay texture. Soil sample number 8 corresponds to

a sandy clay; whereas, samples 2, 7, 9 and 13 are a clay loam. The rest of the soil samples (1, 3, 4, 5, 6, 10, 11, 12, 14, 15, 16, 17 and 18) were sandy clay loam. From **Figure 4**, it is clear that the sandy clay sample 8 has the higher K_s ; whereas, the clay loam (samples 2, 7, 9, 13) is as random as the sandy clay loam. The average K_s was $1.20 \times 10^{-6} \text{ m} \cdot \text{s}^{-1}$ with a standard deviation of 9.95×10^{-7} $\text{m} \cdot \text{s}^{-1}$. The high variation could be the result that these soil samples were from a plowed layer and invariably some compression occurred during sampling confirming that K_s is problematic to measure due to its associated variability and uncertainty [21].

3.2. Rosetta Simulations

Results from comparing measured values of K_s with those obtained with the five levels of Rosetta are given in **Table 4** and plotted in **Figure 5**. As previously, des-



Figure 5. Comparison of measured and calculated values of saturated hydraulic conductivity (K_s , m·s⁻¹) for 18 locations of a field in Lamesa, Texas. Calculated values of K_s were obtained with the Rosetta PTF [14]. (a) Textural class, level H1; (b) H1 plus sand, silt and clay, level H2; (c) H2 plus soil dry-bulk density, level H3; (d) H3 plus soil water content at -33 kPa, level H4; and (e) H4 plus soil water content at -1500 kPa, level H5.

cribed, Rosetta uses a hierarchical approach and each level H is based on the previous one plus an additional surrogate measurement. Five simulations were run with Rosetta and the results for each level of H are plotted in **Figures 5(a)-(e)**.

The input data used in all Rosetta simulations is given in Table 5. These results show that level H1 Rosetta. which calculates K_s using USDA textural class input does not follow the pattern of the measured values (Figure 5(a)). The RMSD between measured and calculated values of K_s obtained in this case was $1.03 \times 10^{-6} \text{ m} \cdot \text{s}^{-1}$ and the NSE was -0.10. Both statistical parameters suggest that the values of K_s obtained with level H1 of Rosetta were wrong for this particular field and this result was confirmed with the low value of R^2 (Figure 6(a)). When the percentage of sand, silt and clay is used, *i.e.* level H2 of Rosetta, the RMSD between measured and calculated values of K_s was 1.15×10^{-6} , NSE was -0.3 and R² was 0.22 (Figure 6(b)), slightly better than values obtained with level H1, but also incorrect. Adding the soil drybulk density, $\rho_{\rm b}$, to the Rosetta input, *i.e.* level H3, shows some improvement with an RMSD of $9.8 \times 10^{-7} \text{ m} \cdot \text{s}^{-1}$ and NSE is 0 with an $R^2 = 0.15$ (Figure 6(c)). For the next level of Rosetta, H4 that uses textural information, ρ_b and soil volumetric water content, θ , at h = -33 kPa as input, the RMSD was 8.63×10^{-7} m·s⁻¹, the NSE is 0.22 and R² = 0.55 (**Figure 6(d)**). In H4, the variability of the measured values of K_s (from 2.3×10^{-7} to 2.7×10^{-6} m·s⁻¹ for sandy clay loam) is larger than the RMSD. Finally, using level H5 of Rosetta, which requires input values of sand, silt and clay, ρ_b , and θ at h = -33 and -1500 kPa, yields an RMSD of 7.81×10^{-7} m·s⁻¹, NSE of 0.36, and R² = 0.51 (**Figure 6(e)**). This level of Rosetta calculated values of K_s with the smallest RMSD and largest NSE.

4. Summary and Conclusions

All water flow and solute transport modeling in soils need as crucial input the K_s [3]. Also K_s is a sensitive parameter in hydrological models and one of the most problematic measurements at field-scale in regard to variability and uncertainty [21]. This study tested the application of the pedotransfer function Rosetta to identify the level of input needed in order to use Rosetta as a tool to calculate the K_s for a 64 ha field in the Southern High Plains of Texas.

Rosetta calculations of K_s improved with a RMSD of

Location Sand % Silt % θ – 1500 kPa Clay % θ – 33 kPa $\rho_{\rm b} \,({\rm kg}\cdot{\rm m}^{-3})$ 1 59.2 15.2 25.6 1680 0.29 0.06 2 44.9 24.6 30.5 1580 0.30 0.05 3 22.9 26.0 1670 0.29 0.13 51.1 0.29 4 53.1 17.5 29.4 1780 0.06 5 46.5 19.8 33.7 1540 0.28 0.14 49.0 0.29 6 17.5 33.6 1510 0.14 7 32.0 47.8 20.2 1710 0.30 0.17 8 46.9 18.4 34.7 1440 0.27 0.14 9 46.5 19.8 33.7 1600 0.32 0.16 10 52.0 16.4 31.6 1600 0.30 0.23 11 49.4 20.1 30.5 1660 0.28 0.15 12 45.5 21.5 32.9 1790 0.29 0.17 44.9 13 21.7 33.4 1580 0.27 0.18 14 54.3 20.2 25.4 1690 0.20 0.13 15 53.1 17.5 29.4 1480 0.29 0.15 16 50.7 17.5 31.8 1720 0.29 0.11 17 52.1 16.4 31.5 1630 0.27 0.15 18 59.2 15.2 25.6 1680 0.29 0.06

Table 5. Measured values of sand, silt, clay, dry-bulk density (ρ_d , kg·m⁻³), and soil volumetric water content (θ) at a suction of -33 kPa and -1500 kPa for the 18 locations shown in Figure 2.



Figure 6. Linear regression between calculated and measured values of saturated hydraulic conductivity (K_s , m·s⁻¹). Calculated values of K_s were obtained with five hierarchical levels of the Rosetta PTF. (a) Textural class, level H1; (b) H1 plus sand, silt and clay, level H2; (c) H2 plus soil dry-bulk density, level H3; (d) H3 plus soil water content at -33 kPa, level H4; and (e) H4 plus soil water content at -1500 kPa, level H5. The line plotted is the 1:1 and the linear regression coefficient (R^2) for each level of H is shown in the upper left hand corner.

 $1.03 \times 10^{-6} \text{ m} \cdot \text{s}^{-1}$ and NSE -0.1 when only using the USDA textural class as input to a RMSD of 9.8×10^{-7} m·s⁻¹ and NSE of 0, when the ρ_b was added. The NSE indicates that with the textural information only, the measured mean value of K_s is a better predictor than the model. However, when the ρ_b was added Rosetta calculations were as accurate as the mean of the measured values. These values were further improved when θ at h = -33 kPa was added as input; the RMSD was reduced to $8.63 \times 10^{-7} \text{ m} \cdot \text{s}^{-1}$ and the NSE increased to 0.2. Furthermore, with the addition of the soil volumetric water content, θ , at h = -1500 kPa, the RMSD decreased to 7.81 ×

 10^{-7} m·s⁻¹ and the NSE increased to 0.36. Once θ information is introduced the NSE is positive, which means that Rosetta does better predictions that the mean of the measured values. Based on these results we conclude that Rosetta PTF is a useful tool that can be used to calculate K_s in the absence of measured values and that for this particular soil, the hierarchical level 5 of Rosetta yielded the best results with the measured input data. The better prediction of K_s by using the hierarchical level 5 of the PTF Rosetta confirms then findings of Gülser *et al.* [20] who used the same input parameters to calculate saturated hydraulic conductivity, K_s .

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