# Non-Linear Regression for Identifying the Optimal Soil Hydraulic Model Parameters

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Abstract. Study is focussed on determination of soil hydraulic parameters and analysis of the moisture retention function for different soil textures. Experiments utilizing pressure plate apparatus were conducted to estimate actual soil moisture characteristics. Optimal values of soil hydraulic parameters of Brooks-Corey (1964), Van-Genuchten (1980) and Modified Van Genuchten (2005) models were estimated using nonlinear regression in SPSS. These parameters were used as input in HYDRUS-1D forward simulation to yield analytical soil moisture retention curve. The analytically obtained moisture retention curve is compared with actual SMC curve to assess the performance of nonlinear regression based approach for identifying optimal soil hydraulic parameter values.

**Keywords:** Soil moisture; Matric potential; Iteration; Unsaturated zone; Optimization.

### 1 Introduction

Water transport in saturated soils takes place as liquid flow and in unsaturated soils as liquid-vapour flow [1]. The soil water flow is generally specified by solving the proper governing partial differential equations for different initial and boundary conditions [2,3]. The soil hydraulic parameters at the soil surface and in the unsaturated zone are key variables, because they are essential inputs to large scale hydro-climatic and hydrologic processes [4]. An important problem commonly encountered in computer based numerical solutions is obtaining the primary input parameters, that is, soil matric potential and hydraulic conductivity as a function of soil water content [5]. Soil water retention functions describes the ability of soils in storing and releasing water and forms an essential parameter in the unsaturated flow and infiltration models [6,7]. Water retention function is dependent on particle-size distribution, organic matter content, mineralogy of clay and hysteresis.

The concept of SMC was initially proposed by [8]. For many years, SMC has been subjected to substantial research and numerous laboratory, field and theoretical techniques are developed for its determination [9-11]. The soil moisture characteristic (SMC) expresses the functional relationship between the suction head  $(\psi)$  and volumetric moisture content  $(\theta)$  in an unsaturated soil [12,13].

Modelling SMC is generally based on solution of governing mathematical equations containing various parameters describing the system. The most popular analytical models available in literature are Brooks-Corey [14], Campbell [15], Van-Genuchten [16], Kosugi [17] and Modified Van-Genuchten [18]. In order to have a reliable model, the estimated value of the parameters should fit their actual ones. The parameters are obtained either experimentally or analytically. In present study, optimal values of parameters of three analytical models [14,16,18] are estimated using nonlinear regression. Following objectives have been set:

- i. Estimation of field parameters of SMC through laboratory experiments and development of experimental SMC using pressure plate apparatus.
- Calculation of optimal soil hydraulic model parameters through nonlinear regression using SPSS.
- iii. Comparison between experimental SMC and model predicted SMC using HYDRUS-1D.

## 2 Material and Methods

## 2.1 Analytical Models

Three analytical models are considered in the study as given below, Brooks and Corey (BC):

$$\frac{\theta - \theta_r}{\theta_s - \theta_r} = \left(\frac{\psi_b}{\psi}\right)^{\lambda} \qquad \text{for } \psi \le \psi_b$$
 (1)

Van Genuchten (VG):

$$\frac{\theta - \theta_r}{\theta_s - \theta_r} = \left[ \frac{1}{1 + \|\alpha_v \psi\|^{n_v}} \right]^m \quad \text{for } \psi \le 0$$
 (2)

Modified Van Genuchten (MVG)

$$\frac{\theta_m - \theta_r}{\theta_s - \theta_r} = \left[ \frac{1}{1 + \left\| \alpha_{mv} \psi \right\|^{n_{mv}}} \right]^m \qquad \text{for } \psi \le 0$$
 (3)

Where,  $\theta_m = \theta_r + (\theta_s - \theta_r)(1 + \|\alpha h_s\|^n)^m$ 

In above models,  $\psi$  is the soil matric potential,  $\psi_b$  is the bubbling pressure,  $\lambda$  is the pore size index,  $\alpha_v$  and  $n_v$  are unsaturated soil parameters with m=1- $(1/n_v)$ ,  $\theta$  is the actual volumetric moisture content (cm<sup>3</sup> cm<sup>-3</sup>) and the subscripts s and r represent saturation and residual values of the moisture content.

## 2.2 Experimental Data

Laboratory experiments were utilized to estimate relevant soil textural and hydraulic parameters. Three different sites were selected and 30 samples from each site were subjected to sieve and hydrometer analysis [19]. The soil texture comes out to be loam, sandy loam and loamy sand from site 1, 2 and 3 respectively. Figure 1 shows the particle size distribution and soil textural classification chart (USDA) for loam

soil. Experimental SMC was estimated using pressure plate apparatus (PPA), which involved concurrent measurements of soil moisture and soil suction. The duration of experiments varied for each sample and for different pressure readings.  $\theta_s$  is assumed to be equal to soil porosity and  $\theta_r$  is estimated from PPA.

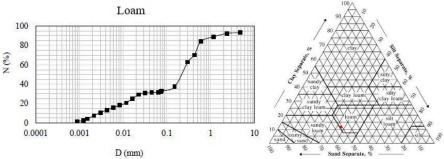


Fig. 1. (a) Particle size distribution (b) Soil textural classification chart for Loam soil

#### 2.3 Nonlinear Regression

The optimal values of model parameters are determined through nonlinear regression using IBM SPSS (Statistical Package for Social Sciences) software. In nonlinear regression, a function which is nonlinear combination of model parameters is used for modelling the data. The function is dependent on one or more independent variables. The experimentally obtained input and output variables are used. An initial guess of parameters is provided, and the iterations are performed. After a certain number of model and derivative assessments, the iteration stops resulting in model fit parameter values.

#### 2.4 HYDRUS

The parameters of the analytical models obtained from SPSS and those from laboratory experiments are used as input data for forward simulation in HYDRUS-1D. HYDRUS is based on soil moisture flow equation [20]. The simulated results in development of analytical SMC from each model considered in the study.

#### 3 Results and Discussion

#### 3.1 Soil Moisture Characteristics

Numerous experiments were conducted using PPA for different soil textures. Thirty set of sample observations were taken for each soil texture to accurately determine its SMC curve. The SMC curves obtained for different soil textures considered in the study are shown in Fig. 2. There is no clear difference between the moisture retention near the saturation end, however at the drier end, the loam exhibits higher moisture retention in comparison to sandy loam and loamy sand soil.

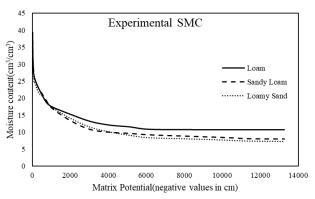


Fig. 2. Experimental soil moisture characteristics curve for different soils

## 3.2 Soil Hydraulic Parameter Estimation

The soil hydraulic parameters of three analytical models is estimated through iterative optimization using nonlinear regression. The optimal values of these parameters for loam, sandy loam and loamy sand is given in Table 1. The parameters are found to be different for each soil texture. The statistical results for nonlinear regression are shown in Table 2. For each model, a number of model assessments (MA) and derivative assessments (DA) are run which converges to a final value as mentioned in Table 2. The R-squared (R²) value of analysis of variance is also given in Table 2. R² is found to be above 0.95 for each model in case of all soil textures considered in the study which suggests that the estimated values are reliable.

Table 1. Values of soil hydraulic model parameters

	SMC models						
Soil Texture	Brooks and Corey		Van Genuchten		Modified Van Genuchten		
	$\psi_b$ (cm)	λ	$\alpha_{\rm v} ({\rm cm}^{-1})$	$n_{\rm v}$	$\alpha_{\rm mv} ({\rm cm}^{-1})$	$n_{\mathrm{mv}}$	
Loam	8.952	0.28	0.032	1.405	0.053	1.402	
Sandy Loam	16.522	0.467	0.048	1.682	0.051	1.663	
Loamy Sand	11.831	0.278	0.017	1.457	0.024	1.362	

Table 2. Statistical results for nonlinear regression

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	SMC models								
Soil Texture	Brool	Brooks and Corey		Van Genuchten			Modified Van Genuchten		
	MA	DA	$\mathbb{R}^2$	MA	DA	$\mathbb{R}^2$	MA	DA	$\mathbb{R}^2$
Loam	12	6	0.97	16	8	0.95	16	8	0.95
Sandy Loam	12	6	0.96	16	8	0.95	14	7	0.96
Loamy Sand	14	7	0.96	16	8	0.96	14	7	0.96

## 3.3 SMC Comparison

The analytical SMC is computed using HYDRUS-1D employing the equations mentioned in 2.1. Each soil texture exhibited its peculiar SMC curve as shown in Figs. 3-5. The graphical comparison between experimental SMC and analytical SMC obtained from three models is shown in Figs. 3-5. From graphical comparison it is found that Van Genuchten presented precise SMC to experimental one. In case of sandy loam soil, all the three models were found to present close agreement with experimental SMC curve. These observations are further substantiated by results of statistical analysis using coefficient of determination (COD) and coefficient of variation (COV). The numerical formulation of COD and COV can be referred in [Shankar et al. 2012].

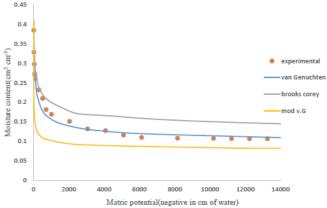


Fig. 3. Comparison of experimental and analytical SMC for loam soil

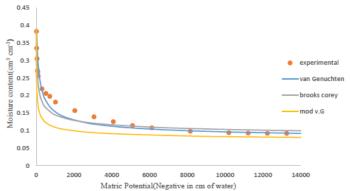


Fig. 4. Comparison of experimental and analytical SMC for sandy loam soil

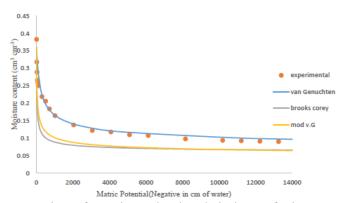


Fig. 5. Comparison of experimental and analytical SMC for loamy sand soil

**Table 3.** Results of statistical comparison between experimental and analytical SMC

	SMC models						
Soil Texture	Brooks and Corey		Van Ger	nuchten	Modified Van Genuchten		
	COD	COV	COD	COV	COD	COV	
Loam	0.812	0.19	0.924	0.06	0.824	0.22	
Sandy Loam	0.884	0.05	0.902	0.04	0.801	0.12	
Loamy Sand	0.786	0.21	0.967	0.02	0.795	0.20	

## 4 Conclusion

Following conclusions are drawn from the study:

- i. Soil hydraulic parameter estimation using nonlinear regression is a reliable method for obtaining optimal values.
- Van Genuchten model simulated SMC was closest to experimental SMC indicating the accuracy of model among other models considered in the study.
- iii. In case of sandy loam soil, all the models presented satisfactory results.

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