

Evaluation of soil texture determination using soil fraction data resulting from laser diffraction method**

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Abstract. There are global aspirations to harmonize soil particle-size distribution data measured by the laser diffraction method and by traditional sedimentation techniques, e.g. sieve-pipette methods. The need has arisen therefore to build up a database, containing particle-size distribution values measured by the sieving and pipette method according to the Hungarian standard (sieve-pipette methods-MSZ) and the laser diffraction method according to a widespread and widely used procedure. In our current publication, 155 soil samples measured with sieve-pipette methods-MSZ and laser diffraction method (Malvern Mastersizer 2000, HydroG dispersion unit) were compared. Through the application of the usual size limits at the laser diffraction method, the clay fraction was under- and the silt fraction was overestimated compared to the sieve-pipette methods-MSZ results, and subsequently the soil texture classes were determined according to the results of both methods also differed significantly from each other. Based on our previous experience, the extension of the upper size limit of the clay fraction from 2 to 7 μm increases the comparability of sieve-pipette methods-MSZ and laser diffraction method, in this way the texture classes derived from the particle-size distributions were also more in accordance with each other. The difference between the results of the two kinds of particle-size distribution measurement methods could be further reduced with the pedotransfer functions presented.

Keywords: laser diffraction, particle-size distribution, pedotransfer function, soil texture triangle

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INTRODUCTION

The accurate and precise analysis of particle-size distribution (PSD) is important in soil science, geology and sedimentology (Miller and Schaetzl, 2011). Most frequently, 0.25-2 mm particles are separated by wet sieving. The under 0.25 mm separation of different-sized particles is based on sedimentation, applying either the pipette or hydrometer method (Gee and Bauder, 1986; Allen, 1990; ISO 11277:2009(E)).

Stokes' law is applied to calculate how much time is needed for a particle with a given diameter to settle in a suspension. In settlement, it is assumed that: (a) soil particles are rigid, spherical and smooth, (b) soil particles have similar densities, (c) particle-to-particle interference and boundary effects from the walls of the sedimentation column are negligible, (d) particle sizes are small enough to ensure that the induced fluid flow is within the laminar flow regime. When particles have irregular shape then deviations from the Stokes' equation are expected, which is characteristic of silty and clayey particles (Clifton *et al.*, 1999).

The laser diffraction method (LDM) for particle-size analysis was introduced by Allen (1990). LDM is being applied more and more often in soil science for soil PSD determination (Bieganski *et al.*, 2018); however, the labour-intensive pipette or hydrometer methods are the ones prescribed in internationally accepted standards for

soil PSD analysis and therefore they are still the most widely used methods. LDM calculates the PSD according to the laser beam diffraction angle and intensity using different models (de Boer *et al.*, 1987). LDM provides the PSD in volume % which is identical with weight % if the shape of the particles are considered to be spherical and they have similar densities (Bieganowski *et al.*, 2012). One of the advantages of LDM analysis is that the PSD of a large number of samples can be measured in less time than with other currently used methods. Laser diffractometers, which have recently been made available, provide a relatively wide measurement range – from fractions of micrometers to several millimeters. Several small categories (bins) can be measured, which serves to provide almost continuous PSD data (Miller and Schaetzl, 2011). The disadvantage of LDM is its higher cost than other PSD measurement techniques (Ryzak and Bieganowski, 2011). The precision of the analysis is influenced by sample preparation, pre-treatment, the optical properties of the particles, parameter settings of the equipment and type of theory used for the evaluation (Ryzak and Bieganowski, 2011; Sochan *et al.*, 2012; Madarász *et al.*, 2012; Bieganowski *et al.*, 2018).

Several papers compare the LDM with the sieve-pipette method (SPM) (*e.g.* Konert and Vandenberghe, 1997; Yang *et al.*, 2015; Makó *et al.*, 2017) and/or the hydrometer method (*e.g.* Di Stefano *et al.*, 2010; Ryzak and Bieganowski, 2010; Orzechowski *et al.*, 2014; Fenton *et al.*, 2015; Fisher *et al.*, 2017). Based on the results of the comparisons LDM tends to underestimate the proportion of the clay fraction and overestimates the silt fraction (Miller and Schaetzl, 2011; Orzechowski *et al.*, 2014; Yang *et al.*, 2015) compared to SPM. Sand fractions measured by LDM and SPM do not differ significantly (Orzechowski *et al.*, 2014). The main reasons for these results, based on the literature, may be summarized as follows: (a) the shape of the clay and silt particles are different from the spherical form, (b) the estimated optical indices of the soil particles are incorrect, (c) the dispersion of soil aggregates are insufficient, (d) the representativeness of all fractions are inappropriate (*e.g.* too many or too few sand particles in the suspension to be measured) (Konert and Vandenberghe, 1997; Fedotov *et al.*, 2007; Polakowski *et al.*, 2014; Bieganowski *et al.*, 2018). The differences obtained using SPM and LDM may have a significant impact on soil texture classifications (Sochan *et al.*, 2012).

Several authors have attempted to change the range of the LDM fraction sizes to match them to the results obtained by SPM. First Konert and Vandenberghe (1997) ascertained that the $<2 \mu\text{m}$ clay content measured using SPM is equivalent to the $<8 \mu\text{m}$ fraction in the case of LDM measurements. This $8 \mu\text{m}$ clay/silt fraction boundary was also found by Fenton *et al.* (2015). During the examination of different types of clays Pabst *et al.* (2000) found an LDM clay/silt size range boundary of $3\text{--}5 \mu\text{m}$ to be the most accurate in comparison with SPM. According to

Buurman *et al.* (2001) the optimum LDM clay/silt fraction boundary may vary from 2 to $8 \mu\text{m}$ depending on the type of sediment. Vandecasteele and De Vos (2003) found on a Belgian soil database that the $0\text{--}6 \mu\text{m}$ LDM clay fraction showed the best agreement with the $0\text{--}2 \mu\text{m}$ clay content recorded using SPM. Fisher *et al.* (2017) found $9 \mu\text{m}$ to be the most applicable value for the clay-silt boundary. In the case of calcareous soils a value of $4 \mu\text{m}$ was recommended by Kerry *et al.* (2009) as the upper limit of the LDM clay fraction. During an SPM-LDM comparative analysis of a representative European soil database the clay/silt fraction boundary was given as $6.6 \mu\text{m}$ for samples with organic matter (OM), and $5.8 \mu\text{m}$ for those samples where OM was removed (Makó *et al.*, 2017). For dune samples, Sitzia *et al.* (2017) found that the LDM clay fraction under $4.6 \mu\text{m}$ showed a good correlation with SPM clay content. It is rarely, it is advised to also change the silt/sand boundary. Fisher *et al.* (2017) suggested changing the upper silt boundary from $20 \mu\text{m}$ to $26 \mu\text{m}$ to obtain a better correlation between the results of the hydrometer methods and LDM. Makó *et al.* (2017) proposed $60.3 \mu\text{m}$ for samples with OM, and $69.2 \mu\text{m}$ for samples without OM as the silt/sand boundary.

Several authors have made an effort to create models for converting LDM PSD data to SPM PSD data (Di Stefano *et al.*, 2010; Yang *et al.*, 2015). Most of these conversion equations, however, were based on small local soil datasets and may not be applied on a national or continental scale. Recently, specific pedotransfer functions (PTFs) with different input requirements were developed by Makó *et al.* (2017) for continental scale applications in Europe to convert data from LDM to SPM.

The aims of the present paper are to 1) present an optimized conversion technique between the PSD data measured according to two well-defined methods (SPM and LDM), 2) show how the accuracy of the conversion method may be verified, with special regard for the accuracy of the classifications into texture classes based on the PSD results obtained with different methods.

MATERIALS AND METHODS

A comprehensive soil survey campaign was carried out in 2014 and 2015 in the Tokaj Wine Region, North-Hungary (Fig. 1). The Tokaj Region is one of the most famous wine regions in Hungary, the vineyards cover the foothills of the volcanic Tokaj Mountains (the elevation goes up to 700 m , Baltic). As a result of both natural and human-induced landscape evolution, the Tokaj Region has a spatially diverse and unique terroir both in the pedological-geological and climatic sense. The differently weathered rhyolite tuff and windblown loess are common parent materials. In the region of the rhyolite parent material, the combined effect of the progressive alteration of tuff and the slope position results in diverse soil sequences. On the elevated position,

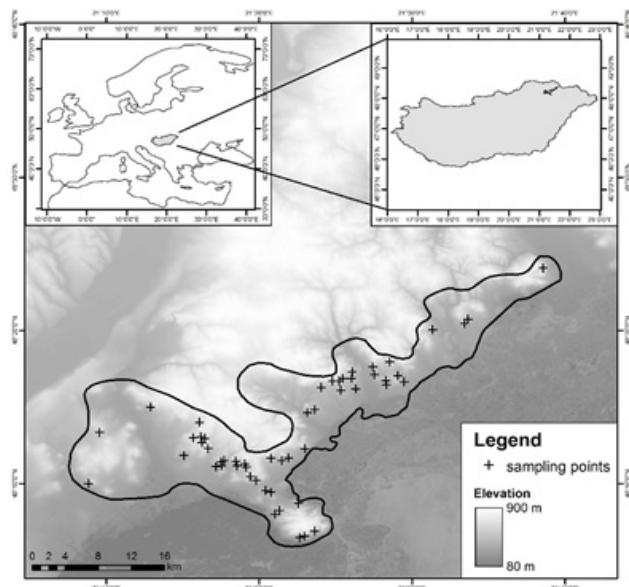


Fig. 1. Map of soil sampling points.

where the erosion is well expressed, the fresh, unaltered, but usually substantially fractured rhyolite tuff may be present at a shallow depth. Cambic Leptosols dominate these sites, alternating with Lithic or Mollic Leptosols in some places. On the lower slopes we find mainly physically weathered volcanic “sand” in which the easily altered feldspar minerals have entirely transformed to clay, while some of the mafic minerals remained observable in the samples (mostly biotite). As a result of progressive chemical alteration and mass-transport processes, fine, altered, clayey material covers most of the surfaces. On loess-covered surfaces (mainly in the Southern part of the pilot area) mass transport processes are dominant, resulting in Calcisols with a thin humic upper layer. The secondary carbonate accumulation in the deeper layers is predictable in the loess area, but – because of the carbonate remains being present in the previously formed, probably most extended loess cover – it appears as a common phenomenon in the rhyolite tuff area also, at about a 1.5 m depth. Based on the soil survey database, 53 locations were selected for further investigation, in an effort to represent the local variability both in terms of the terrain position and the parent material. The bulk samples were taken from boreholes with a depth of approximately 2 m, depending on the layering of the profiles. 155 soil samples were collected from the 53 soil profiles (Fig. 1).

The conventional sieve-pipette method was used according to the Hungarian standard (MSZ-08. 0205: 1978) (SPM-MSZ) and the laser diffraction method (LDM) was also used to measure the particle-size distribution (PSD) of the soils. Before PSD analysis for both the SPM-MSZ and the LDM, the soil samples were air-dried, gently crushed, and dry-sieved with a 2000 μm mesh size. Macroscopic traces of organic matter (roots, chaffs, debris *etc.*) were physically removed. Prior to the SPM-MSZ and LDM analy-

sis neither organic matter, nor carbonates and iron oxides were removed, as recommended by the Hungarian standard. The only chemical pre-treatment of the soil samples before PSD analysis was the addition of dispersing agents (see later).

SPM-MSZ procedure: The PSD was determined by a combination of sieving and sedimentation, starting from a 25 g sample of air-dried soil. Particles smaller than 2000 μm and retained on a 250 μm aperture sieve (coarse sand fraction) were determined by a wet sieving procedure. Particles, which passed through the latter sieve were determined by sedimentation using the pipette method. For the calculation of sedimentation time it was assumed that the mean particle density is 2.65 Mg m^{-3} . The method does not require pre-treatments (organic matter, CaCO_3 or iron oxyhydroxides removal) before the measurements take place. The chemical dispersion was performed using sodium pyrophosphate (55.8 g L^{-1}). The physical method used to facilitate the dispersion was shaking the pre-treated suspension for 6–10 h on the end-over-end shaker. By performing the PSD measurements according to the MSZ standard, the percentage (mass %) of the constituent fractions (SPM-MSZ_sand: 2000 μm to 50 μm ; SPM-MSZ_silt: 50 to 2 μm ; SPM-MSZ_clay: < 2 μm) could be calculated.

LDM procedure: For LDM analysis the Mastersizer 2000 (Malvern Company, UK) laser diffractometer was used (ISO 13320:2009). This instrument allows for the measurement of the PSD (volume, %) within the size range of 0.02–2000 μm (Malvern Operators Guide, 1999). The measurements were conducted using a Hydro 2000G dispersion unit. The PSD measurements using LDM were usually repeated twice. Third or fourth repetitions were measured, when the two previous repetitions were significantly different. The mass of the dry soil samples placed into the dispersion unit were included in the range of 0.5 to 1 g depending on the obscuration of the soil suspension after dispersion. According to the manual of the Mastersizer 2000 the obscuration values had to always be between 10 and 20%. The dry soil sample was dampened by the dropwise addition of standard Calgon (33 g of sodium hexametaphosphate and 7 g of anhydrous sodium carbonate in water to make 1 L of solution) dispersant on a watch glass. Thereafter the paste was washed into the tank of the dispersion unit and a further 25 cm^3 Calgon solution was poured on it. The measurements were carried out in a $\sim 800 \text{ cm}^3$ water suspension using deionized water (Bieganski *et al.*, 2010). In order to ensure complete disaggregation and dispersion the soil suspension was then treated for 240 s with ultrasound at 75 % of maximum power (applying 0.75 x 35 W and 0.75 x 40 kHz). The Hydro 2000G dispersion unit pump speed was set at 1750 r.p.m. and the stirrer at 700 r.p.m. (Sochan *et al.*, 2012). Using this setting, we could achieve the maximum homogenization of the suspension in the beaker while eliminating air bubbles (Ryzak and Bieganski, 2011). The intensity of the laser light

registered with the particular detectors of the measurement system was converted to PSD according to the Mie theory, and the following optical setting parameters were used: a refraction index of 1.52 and an absorption index of 0.1 for the dispersed phase, and a refraction index of 1.33 for water as the dispersing phase. The laser light wavelength in the apparatus was 466 nm for blue and 633 nm for red light. All of the measurements lasted for 60 s (30 s for blue and 30 s for red light) (Ryżak and Bieganowski, 2010). The selected algorithms of the PSD calculation were the following: 1) general purpose analysis (GPA), 2) irregular shape ratio (ISR) (Malvern Operators Guide, 1999).

A dataset ($N = 325$) was built from the PSD measurements, which includes besides the SPM-MSZ and LDM PSD data the results of basic soil analyses and a description of the soil samples. The basic soil properties were measured according to the Hungarian standards (Buzás, 1993). To obtain LDM particle size fractions similar to those of the pipette method, first the nearest fraction boundaries were chosen, and the fractions 1.90 and 52.5 μm – values closest to 2 and 50 μm – were used from the distribution data (“original LDM fraction boundaries”). The original LDM PSD fractions were accordingly LDM_sand: 2000 to 52.5 μm , LDM_silt: 52.5 to 1.90 μm , LDM_clay: < 1.9 μm .

Fourteen possible clay fractions were calculated from the PSD results obtained by LDM using an upper boundary ranging from 1.4 to 10.0 μm (1.4, 1.6, 1.9, 2.2, 2.9, 3.3, 3.8, 4.4, 5.0, 5.8, 6.6, 7.6, 8.7 and 10.0 μm , respectively). In the same way, 12 possible sand fractions were considered using a lower boundary ranging from 30.2 to 138.4 μm (30.2, 34.7, 39.8, 45.7, 52.5, 60.3, 69.2, 79.4, 91.2, 104.7, 120.2 and 138.2 μm , respectively). In order to optimize the thresholds, close to the optimal clay/silt boundary (between 6.6 and 7.6 μm) and optimal silt/sand boundary (between 45.7 and 52.5 μm) different fraction boundaries were examined using linear interpolation with a resolution 0.1 μm . The calculated clay, silt and sand fractions with various boundaries were used as dependent variables against the conventional clay, silt and sand fraction measured with SPM-MSZ (<2.0 μm , 2-50 μm and 50-2000 μm , correspondingly). Lin’s concordance correlation coefficient (CCC) computation (Lin, 1989) between the PSD results of SPM and LDM data for each dataset were performed to optimize the clay, silt and sand fraction boundaries of the LDM measurements. Lin’s CCC measurements indicate how well a new set of observations (LDM PSD measurements) reproduce an original set (SPM-MSZ PSD), so that it may be used to assess the effectiveness of the laser measurement methods (Fisher *et al.*, 2017; Makó *et al.*, 2017). Lin’s CCC indicates perfect agreement if its value is 1, and the evaluation criteria can be divided into the following categories: 0.21-0.40 fair; 0.41-0.60 moderate; 0.61-0.80 substantial and 0.81-1.00 almost perfect. The resulting “modified opti-

num LDM fraction boundaries” were used to calculate the modified clay, silt and sand fractions (LDM_clay_{mod}, LDM_silt_{mod} and LDM_sand_{mod}, respectively).

One of the potential applications of the datasets presented above is to develop methods for calculating conventional (SPM-MSZ) PSD data from more easily measured LDM PSD data and possibly consider other simply measurable soil parameters as well. In order to express the relationship between the results of the SPM-MSZ and LDM, pedotransfer functions (PTFs) were derived. We focused on the prediction method based on input soil data, which are available from LDM measurements and from basic soil tests. A linear regression with a backward elimination method (SPSS Statistics, Regression, Linear regression; SPSS, 2004) was used for the development of PTFs, because accuracy and reliability of that statistical method was found to be appropriate in our previous research (Makó *et al.*, 2017). Multiple linear regression equations were constructed to correlate the PSD fractions – clay and silt separately – of SPM-MSZ (SPM-MSZ_{clay}, SPM-MSZ_{silt} – mass %) with those measured by LDM, using the modified optimum LDM clay and silt fractions (LDM_clay_{mod}, LDM_silt_{mod} – volume %) calculated with the modified optimum LDM fraction boundaries, resulting in the closest relationship with SPM measurements during Lin’s CCC analysis. In order to predict the SPM-MSZ clay fraction (SPM-MSZ_{clay}_{pred}) apart from the LDM_clay_{mod} we also used the following soil properties as regressed variables: the modified optimum LDM sand content (LDM_sand_{mod} – volume %), humus (mass %) and calcium-carbonate (CaCO₃ – mass %) content, pH(H₂O). The prediction of SPM-MSZ_{silt}_{pred} was performed based on LDM_silt_{mod}, LDM_sand_{mod}, humus, CaCO₃ and pH(H₂O). In addition to the above, transformed soil parameters were calculated and linear, quadratic, reciprocal and common base logarithms of the soil properties involved were also used in the regression analysis. The predicted SPM-MSZ sand fraction was in all cases obtained as follows: SPM-MSZ_{sand}_{pred} = 100 – (SPM-MSZ_{clay}_{pred} + SPM-MSZ_{silt}_{pred}), where SPM-MSZ_{sand}_{pred}, SPM-MSZ_{clay}_{pred} and SPM-MSZ_{silt}_{pred} are the predicted SPM-MSZ sand, clay and silt fractions, respectively (mass %). The accuracy of the developed PTFs was investigated first in the traditional way based on the evaluation of the coefficients of determination (R^2) and root mean square errors (RMSE) (mass %):

$$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^N (y_i - \hat{y}_i)^2},$$

where: y_i stands for the SPM-MSZ_{clay}, SPM-MSZ_{silt} or SPM-MSZ_{sand} values, \hat{y}_i is the predicted SPM-MSZ_{clay}_{pred}, SPM-MSZ_{silt}_{pred} or SPM-MSZ_{sand}_{pred}, N is the total number of samples. Thereafter Lin’s CCC computation between the PSD results of SPM-MSZ and predicted SPM-MSZ (SPM_{pred}) data was also performed. The RMSE

and Lin's CCC computations were also performed for comparison between the measured SPM-MSZ and LDM fractions both for the original and the modified fraction boundary values.

The samples of the dataset were categorized into USDA textural classes in R (R Core Team, 2013), with a 'soiltexture' package (Moeys, 2014). Before assignment to these types, the determined sand, silt, and clay fraction percentages were first normalized and if their sum was not equal to 100%, their values were corrected. The textural classes, which were determined from the measured LDM (using different fraction boundaries) and predicted SPM-MSZ PSD data were compared with measured SPM-MSZ data. The percentage correspondence between the textural classes was then evaluated. The differences between the USDA texture classes calculated from predicted and measured SPM-MSZ were visualized in a texture triangle also with the 'soiltexture' package. The two-dimensional kernel density estimation procedure was used in R to visualize and assess the structure of the soil texture data (Moeys, 2014). The kernel density estimation approximates the probability density function of a random variable and it is a popular tool for encoding the density of observations and for plotting the shape of a distribution (Venables and Ripley, 2002). It essentially finds dense regions of points and therefore it is also able to perform density-based clustering, where intra-cluster similarity is maximized and intercluster similarity is minimized.

The particle-size data is compositional (the sum of the particle size fractions is 100%), accordingly an increase in one fraction will certainly lead to a decrease in other fractions. Therefore, we assessed in a new way the correspondence between the different PSD results. This assessment is based on the soil texture triangles. Using the TT.css2xy function of the 'soiltexture' package, the measured and predicted texture data (three fraction data) were converted into x-y coordinates. This allowed for the calculation of the distance between the measured SPM-MSZ and LDM PSD or a measured and predicted SPM-MSZ PSD of each sample in the soil texture triangle. Using this dimensionless 'distance' variable we evaluated the similarity between the soil textures derived from SPM-MSZ and LDM PSD data in cases of selection between the different size fraction boundaries or the correspondence between the soil textures determined from the measured and predicted SPM-MSZ PSD. The significance of difference between the differently measured/derived PSD data was tested and compared the calculated 'distance' parameter with the One-Way ANOVA, Duncan and Tamhane's T2 Post Hoc Tests (SPSS, 2004).

RESULTS AND DISCUSSION

The selected soil samples show a wide range of variation in some soil properties (Table 1).

Table 1. Description statistics of the physical and chemical properties of the soil samples used for the PSD measurements (N = 155)

Soil properties	Mean	Standard deviation	Min	Max
Clay	35.3	12.4	7.5	67.9
Silt	45.6	11.4	12.1	65.4
Sand	19.1	12.4	2.7	78.2
pH(H ₂ O)	7.2	1.0	5.0	8.9
Humus	1.1	0.6	0.2	3.1
CaCO ₃	1.8	4.1	0	36.0

Clay – clay content (<2 mm), Silt – silt content (2-50 mm), Sand – sand content (50-2000 mm), all of these fractions were determined by SPM-MSZ (MSZ-08.0205:1978).

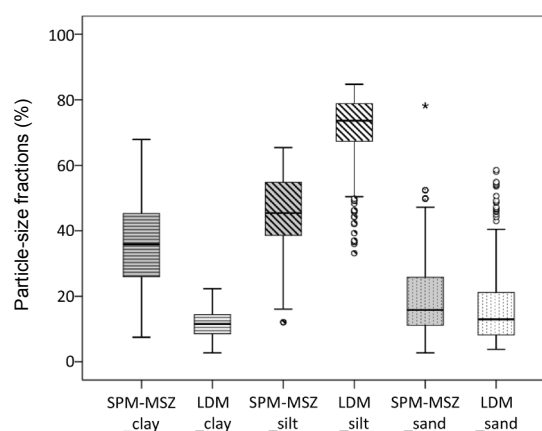


Fig. 2. Comparison of the interquartile ranges of soil particle-size fractions as measured by different PSD measurement methods. SPM-MSZ: PSD from a sieve-pipette according to the Hungarian standard; LDM: PSD from a laser diffractometer using the original fraction boundaries.

Comparing the results of the SPM and LDM measurement methods (boxplots in Fig. 2) the mean clay (2 μm >), silt (2-50 μm) and sand (50 μm <) fractions were 35.3, 45.6 and 19.1% for SPM (based on the MSZ-08.0205:1978. standard particle-size limits), and 11.9, 71.1 and 17.0%, respectively, for LDM where the size ranges of the LDM fractions approximately corresponded to the MSZ size classes of SPM. The results confirmed that the LDM underestimated the clay content relative to SPM (on average the value determined by LDM is only 33% of the SPM-MSZ value), but overestimated the silt content (on average LDM is 156% of the SPM-MSZ). LDM sand was slightly underestimated (on average LDM is 88% of the SPM-MSZ). The measured fractions also had different interquartile ranges (Fig. 2). This corresponds to the results of other authors (Konert and Vandenberghe, 1997; Orzechowski *et al.*, 2014; Yang *et al.*, 2015).

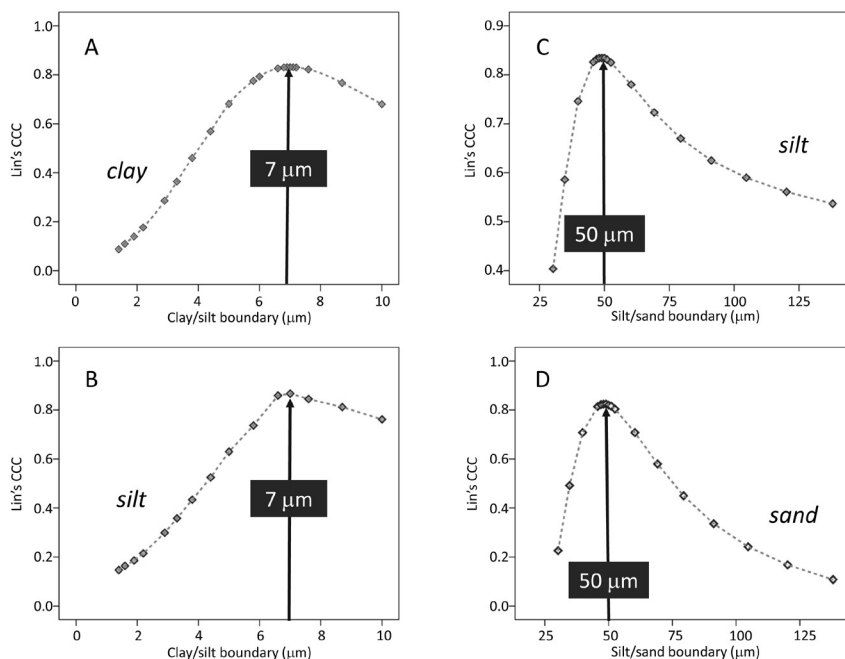


Fig. 3. Lin's concordance correlation coefficient (CCC) values of the LDM's and SPM's (A and B) clay-silt and (C and D) silt-sand fraction boundaries. The y-axis gives Lin's CCC of the (A and B) cumulative size classes of the LDM_clay (from <1.4 to <10.0 μm) and the SPM-MSZ_clay (<2 μm) fraction and (C and D) cumulative size classes of the LDM_sand (from >30.2 to >138.4 μm) and the SPM-MSZ_sand (>50 μm) fraction.

In order to identify the optimal boundaries between the clay, silt and sand fractions obtained with LDM, the statistical method suggested by Lin (1989) seemed to be useful. The results of Lin's concordance correlation coefficient (CCC) analyses are summarized in Fig. 3. The optimized clay-silt fraction boundaries according to the change in Lin's CCC values, and based on a comparison of the clay and silt fractions, are shown in Fig. 3A, B. For the clay fraction, the CCC value was the highest for the 0-7.0 μm fraction (CCC: 0.831) (Fig. 3A). The upper boundary calculated for clay (7.0 μm) is acceptable as the lower boundary of the silt fraction as well (CCC: 0.867), as illustrated in Fig. 3B. This is in line with the findings of several authors, the underestimation of the amount of clay by LDM may be compensated by changing the clay-silt boundary (Konert and Vandenberghe, 1997; Vandecasteele and De Vos, 2003; Fenton *et al.*, 2015; Fisher *et al.*, 2017). The optimal silt/sand boundary is shown in Fig. 3C, D. In Fig. 3C the similarities between the fractions were tested as measured by SPM and LDM when the lower limit was set to 7 μm and the upper boundary was varied between 30.2 and 138.4 μm . The highest correspondence was found at 50 μm both for silt (CCC: 0.834) (Fig. 3C) and sand (CCC: 0.910) (Fig. 3D). The clay, silt and sand fraction (volume %) belong to the 7 and 50 μm size limits, and cannot be evaluated directly from the measured LDM results due to the default settings of the instrument, thus they are calculated from the cumulative values of the next/closer measured point (6.6 and 7.6 μm , and so 45.7 and 52.5 μm) with a lin-

ear interpolation (At the time of the measurements, we did not know the optimal fraction size boundaries, they were only optimized during data processing). Since all of the values of Lin's CCC are above 0.8, the correspondences are considered to be very good.

The PSD derived by the SPM and LDM methods using the modified optimal fraction boundaries (7 and 50 μm) for LDM may be seen in Fig. 4. It may be observed, that

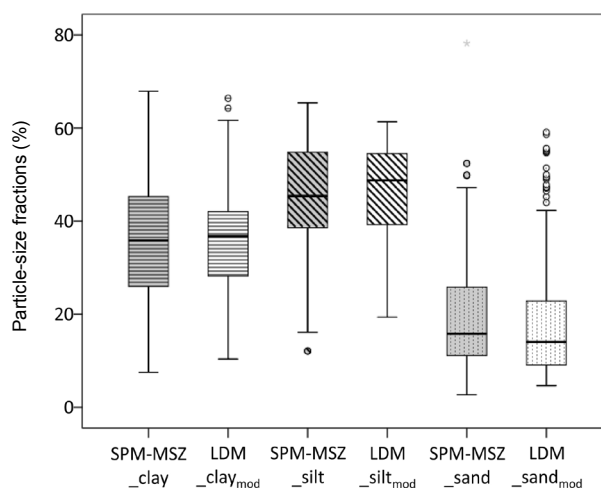


Fig. 4. Comparison of the interquartile ranges of soil particle-size fractions measured by different (SPM-MSZ, sieve-pipette according to Hungarian standard; LDM, laser diffractometer) methods with modified (optimal) PSD boundaries. The 'LDM_clay_mod', 'LDM_silt_mod' and 'LDM_sand_mod' indicate the modified fraction boundaries.

Table 2. Recommended conversion equations (pedotransfer functions)

Conversion model
$\text{SPM-MSZ_clay}_{\text{pred}} = -22.24 + 0.62 \times \text{LDM_sand}_{\text{mod}} + 11.89 \times \sqrt{\text{LDM_clay}_{\text{mod}}} + 111.93 \times 1/\text{pH}(\text{H}_2\text{O}) - 7.77 \times 1/\text{humus} - 25.67 \times \log_{10}\text{LDM_sand}_{\text{mod}} - 27.67 \times \log_{10}\text{humus}$
$\text{SPM-MSZ_silt}_{\text{pred}} = -10.47 - 0.36 \times \text{LDM_sand}_{\text{mod}} + 0.01 \times \text{LDM_silt}_{\text{mod}}^2 + 0.18 \times \text{pH}(\text{H}_2\text{O})^2 + 4.75 \times \sqrt{\text{humus}} + 19.92 \times \log_{10}\text{LDM_sand}_{\text{mod}}$
$\text{SPM-MSZ_sand}_{\text{pred}} = 100 - (\text{SPM-MSZ_clay}_{\text{pred}} + \text{SPM-MSZ_silt}_{\text{pred}})$

the amounts of each fraction determined by the different methods are significantly more similar after the fraction boundary optimization than they were without boundary modification (Fig. 2).

Based on our previous research and experience (Makó *et al.*, 2017), linear regression was used during the development of the conversion models, including transformed (quadratic, reciprocal and logarithmic) variables as independent variables. Table 2 summarizes the recommended PTFs for converting PSDs from LDM (volume %) to SPM data (mass %), if CaCO_3 and humus content, $\text{pH}(\text{H}_2\text{O})$ are also considered to relate the SPM and LDM PSD.

In Fig. 5 we present the process by which the modification of the fraction boundary and the application of PTFs changed the rates of the PSD fractions measured by different methods. It may be seen that through the application of the original boundary, LDM underestimated clay content (Fig. 5A), while it overestimated silt content (Fig. 5B), the sand content accordance was much better between the SPM and LDM methods (Fig. 5C). After modifying the size limits the under- and overestimations were almost eliminated, and the correlation between the results of the two methods improved. In the case of clay the improvement was lower (R^2 increased from 0.69 to 0.72) than it was in the case of silt (R^2 increased from 0.49 to 0.75). The coefficient of determination between PSD measured by SPM and PSD predicted with PTF ($\text{SPM-MSZ_clay}_{\text{pred}}/\text{SPM-MSZ_silt}_{\text{pred}}/\text{SPM-MSZ_sand}_{\text{pred}}$) was based on LDM data and the basic soil properties R^2 values were higher, either for clay (0.80), or for silt (0.82). In case of sand the coefficient of determination barely changed ($R^2 = 0.84 - 0.85$).

In order to evaluate the similarities between the fractions, a different set of measurements were made, according to the *RMSE* values it may be assumed that changing the fraction boundary resulted in a significant improvement (*RMSE* decreased for clay from 25 to 6.5; for silt from 27 to 5.7). The PTF prediction further increased the correspondence (*RMSE* = 5.5 for clay and 4.8 for silt). In the case of sand, either a boundary change, or PTF prediction only caused a small reduction in *RMSE* values (5.5; 5.3; 4.8) (Table 3).

A comparison of PSD fractions determined in different ways was also performed with Lin's CCC method (Table 4). After modifying the fraction boundaries of PSD measured by LDM, the LDM and SPM-MSZ PSD were similar

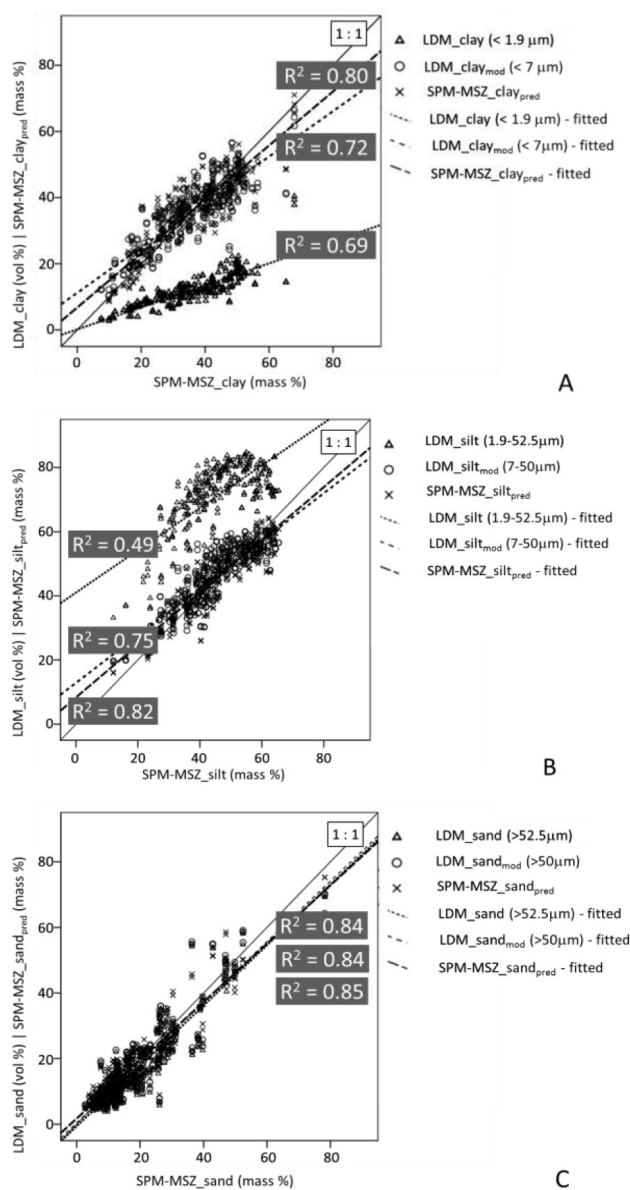


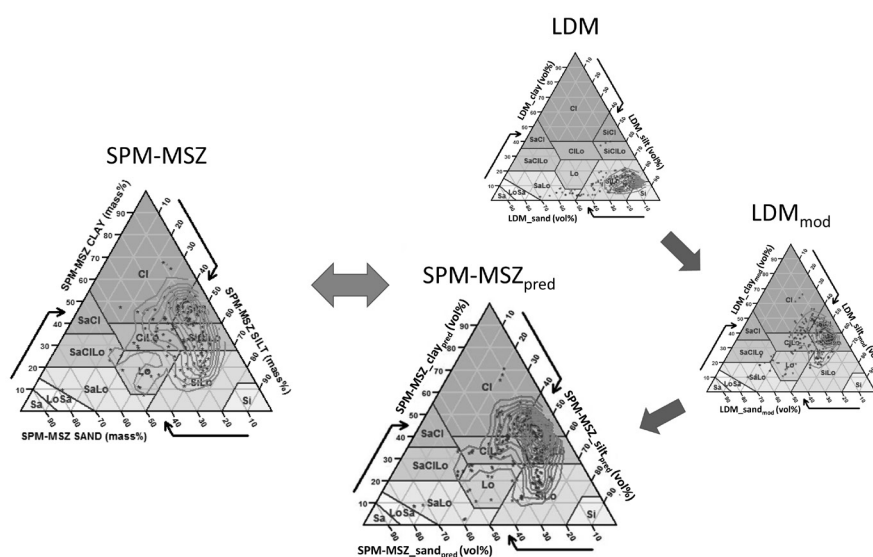
Fig. 5. Evaluation of the efficiency of fraction boundary modification ($\text{LDM_clay}_{\text{mod}}/\text{LDM_silt}_{\text{mod}}/\text{LDM_sand}_{\text{mod}}$) and the use of pedotransfer functions ($\text{SPM-MSZ_clay}_{\text{pred}}/\text{SPM-MSZ_silt}_{\text{pred}}/\text{SPM-MSZ_sand}_{\text{pred}}$) for the conversion of LDM PSD to SPM-MSZ PSD (comparison the A – clay, B – silt, C – sand fractions).

Table 3. Evaluating similarities in the fractions from LDM and SPM-MSZ based on the *RMSE* values

Clay fraction	<i>RMSE</i> (m %)	Silt fraction	<i>RMSE</i> (m %)	Sand fraction	<i>RMSE</i> (m %)
LDM_clay	25.0	LDM_silt	27.0	LDM_sand	5.5
LDM_clay _{mod}	6.5	LDM_silt _{mod}	5.7	LDM_sand _{mod}	5.3
SPM-MSZ_clay _{pred}	5.5	SPM-MSZ_silt _{pred}	4.8	SPM-MSZ_sand _{pred}	4.8

Table 4. Evaluating similarities in the fractions from LDM and SPM-MSZ based on Lin's CCC values

Clay fraction	Lin's CCC	Silt fraction	Lin's CCC	Sand fraction	Lin's CCC
LDM_clay	0.14	LDM_silt	0.19	LDM_sand	0.90
LDM_clay _{mod}	0.83	LDM_silt _{mod}	0.85	LDM_sand _{mod}	0.91
SPM-MSZ_clay _{pred}	0.89	SPM-MSZ_silt _{pred}	0.90	SPM-MSZ_sand _{pred}	0.92

**Fig. 6.** Particle-size data of samples ($N = 325$; each sample represented by a black dot) plotted over the USDA soil texture triangle (Moeys, 2014) according to their % of clay, % of silt and % of sand particles. Abbreviations for the textural classes are: Cl – clay, ClLo – clay loam, L – loam, LoSa – loamy sand; Sa – sand; SaCl – sandy clay, SaClLo – sandy clay loam, SaLo – sandy loam; Si – silt, SiLo – silt loam, SiCl – silty clay, SiClLo – silty clay loam. Contour lines denote the points with the same occurrence frequency.

(Lin's CCC increased for clay from 0.14 to 0.83; for silt from 0.19 to 0.85). When PTFs were also used to transform LDM PSD to SPM-MSZ PSD the similarity further improved (Lin's CCC for clay: 0.89, for silt: 0.90, which means almost perfect agreement). There was no significant Lin's CCC increase for the sand fraction.

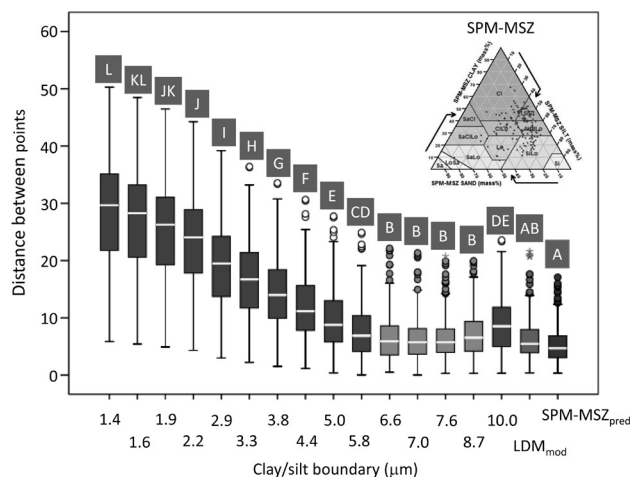
The above results are also shown in texture triangles (Fig. 6). Using the PTF density of PSD, the measured and predicted data are quite similar (SPM-MSZ_{pred} vs. SPM-MSZ). However, the soil texture triangle cannot quantify the accordance of the texture classes between the SPM-MSZ and LDM methods. It may be stated that, although the concordance of the texture classes increased considerably, the agreement did not exceed 63% (Table 5). In this regard, there was no difference between the efficiency of the modified LDM results (LDM_{mod}) and the predicted SPM-MSZ results (SPM-MSZ_{pred}) for any texture classes. At the same

time, it is noteworthy that there are differences in concordance between LDM_{mod} and SPM-MSZ_{pred} for a particular texture class.

The result of the above investigation only provides information on the extent of the agreement in percentage terms between the texture classes defined by different methods. It does not provide information concerning the extent of the soil texture deviations, if they differ. The distance measurement based on the texture triangle may be a suitable method for characterizing this deviation. Figure 7 summarizes the calculated geometric distances of the soil texture 'triangle-texture points' (per sample and method) from each other, or more precisely, the distance between the "texture points" determined based on the results of the SPM-MSZ method from the points, which are derived from the results of the LDM measurements. The boxplot charts show the distance measurement statistics using different

Table 5. Evaluating similarities in the soil textures based on the percentage of concordance between the LDM PSD and SPM-MSZ PSD (An explanation of the abbreviations is given in Fig. 6)

Evaluation methods	Soil texture classes								
	Cl	ClLo	Lo	SaClLo	SaLo	SiLo	SiCl	SiClLo	Total
	Concordance (%)								
LDM	0	0	0	0	100	93	0	0	18
LDM _{mod}	56	56	50	63	100	68	46	90	63
SPM-MSZ _{pred}	53	59	45	38	100	73	61	74	63

**Fig. 7.** Evaluation of the accuracy of the LMD SPM-MSZ conversions achieved by the use fraction boundary changes and PTFs on the basis of texture triangular diagrams (means denoted by the same letter did not significantly differ at $p < 0.05$ according to the multiple range test).

clay/silt fraction boundary switches, optimum modified fraction boundaries (LDM_{mod}, where the clay/silt fraction size limit is $7.0 \mu\text{m}$ and the silt/sand size limit is $50 \mu\text{m}$) and predicted SPM-MSZ PSD results (SPM-MSZ_{pred}). The boxplot charts denoted by uppercase letters show the results of the comparison with One-Way ANOVA, Duncan or Tamhane's T2 Post Hoc Tests. The modification of the LDM soil fraction size boundaries resulted in a much greater improvement in the correspondence of the texture distribution between SPM-MSZ and LDM data, as seen in Fig. 7. The highest degree of accuracy (the smallest deviation) was achieved at the $7 \mu\text{m}$ clay/silt and $50 \mu\text{m}$ silt/sand size limit. The LDM-SPM conversion efficiency may be further improved by the use of PTFs.

CONCLUSIONS

1. Several statistical methods were presented to investigate how the distinct predetermined fraction size boundaries of particles affect the correspondence between the results of sieve-pipette method and laser diffraction method.

2. It was verified that, for a given measurement method and for a given sample material the optimum clay/silt and silt/sand boundaries were 7 and $50 \mu\text{m}$, respectively.

3. It was presented, that the developed pedotransfer functions can further improve the comparability of the laser diffraction method and the sieve-pipette method.

These pedotransfer functions (their accuracy and reliability) were not studied more thoroughly, because they were developed on a relatively small soil sample set and their validity applies only to a sample material with similar properties.

4. The aim of this paper was to highlight that the correspondence between the results of sieve-pipette method according to the Hungarian standards and laser diffraction method, is influenced by other soil properties and that it is possible to take them into account, e.g. with the development of pedotransfer functions.

Conflict of interest: The Authors do not declare conflict of interest.

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