ABSTRACT

The spatial correlation between soil properties and weeds is relevant in agronomic and environmental terms. The analysis of this correlation is crucial for the interpretation of its meaning, for influencing factors such as dispersal mechanisms, seed production and survival, and the range of influence of soil management techniques. This study aimed to evaluate the spatial correlation between the physical properties of soil and weeds in no-tillage (NT) and conventional tillage (CT) systems. The following physical properties of soil and weeds were analyzed: soil bulk density, macroporosity, microporosity, total porosity, aeration capacity of soil matrix, soil water content at field capacity, weed shoot biomass, weed density, *Commelina benghalensis* density, and *Bidens pilosa* density. Generally, the ranges of the spatial correlations were higher in NT than in CT. The cross-variograms showed that many variables have a structure of combined spatial variation and can therefore be mapped from one another by co-kriging. This combined variation also allows inferences about the physical and biological meanings of the study variables. Results also showed that soil management systems influence the spatial dependence structure significantly.

Keywords: cross-variogram, co-kriging, physical and biological meanings.
INTRODUCTION

The spatial correlation between soil and weed variables has attracted the attention of researchers in recent years, due to their agronomic and environmental significance. Andreasen et al. (1991) and Bonham et al. (1995) established the spatial correlation between weeds and soil properties from linear correlation coefficients, without taking spatial positioning into consideration. Furthermore, the co-variation between soil properties and weed distribution is not yet well understood. Thus, understanding the degree of correlation between soil physical properties and the influence of biological factors such as dispersal mechanisms, seed production and survival, and the range of influence of factors such as soil and weed management is crucial for interpreting the meaning of the spatial correlations between these variables (Dieleman et al., 2000).

The management system of a soil is considered one of the most important sources for variability of soil physical properties (Stone and Silveira, 2001; Schaffrath et al., 2008; Vitória et al., 2012). Studies comparing no-tillage and conventional tillage indicate that the main effects of management are an increase in soil bulk density in the top 0.20 m and a decrease in total porosity, especially in relation to the decrease in soil macroporosity and the increase in soil resistance to penetration in no-tillage systems (Tormena et al., 1999a; Stone and Silveira, 2001; Tormena et al., 2004; Falleiro et al., 2003; Sepaskhah et al., 2005). Regarding weed occurrence, the systems are also influential, mainly in terms of the higher weed density and biomass in conventional tillage. In no-tillage, the influence is verified in the larger diversity of the infesting community, with lower biomass (Schaffrath et al., 2007). Therefore, weeds behave similarly to crop plants, which tend to have a lower yield in more compacted soils (Freddi et al., 2008).

Weed mapping from soil properties was proposed by Heisel et al. (1999) since these properties are more easily measurable than plant density. It was observed that the soil silt content correlated well with the density of Lamium spp. Nordmeyer and Dunker (1999) also demonstrated a significant spatial correlation between physical and chemical soil properties and the density of different weed species. Alopecurus myosuroides and Viola arvensis were significantly positively correlated with plant-available magnesium. In contrast, the increase in density in Poa annua correlated with a decrease in the clay amount, available Mg and K content, in total N, and organic C. Walter et al. (2002) studied the spatial correlation between six soil variables and the density of some weed species, verifying that Poa annua and Verónica spp correlated negatively with soil pH and that Viola arvensis correlated negatively with clay content and positively with organic matter content. Other researchers investigated spatial correlations between soil and vegetation variables (Maestre and Cortina, 2002; Holmes et al., 2005; Maestre et al., 2005) and of soil variables with crop yield and soil management over time (Vieira and Dechen, 2010).

Regarding weeds, there is a lack of correlation studies between these and soil physical properties, especially studies which take different soil management systems into consideration. Specifically in Brazil, there is no information in the literature dealing with the possible relationships between soil physical properties and weed species. From these considerations, the hypothesis of this study is that soil physical properties and weed variables are correlated in space and that they can be mapped from one another using co-kriging. To test this hypothesis, cross-variograms between the physical properties of the soil and weeds were established from mathematical models which explain the existing variance structure of these variables. The aim of this study was to evaluate the spatial correlation between soil physical properties and weeds in no-tillage (NT) and conventional tillage (CT) in an Oxisol.
MATERIAL AND METHODS

This study was conducted in an area of the Agro-industrial Cooperative Experimental Farm of Campo Mournão, located in the municipality of Campo Mournão, Northwestern region of Paraná (latitude 23° 17' 09" S, longitude 51° 55' 17" W, 620 m asl). The regional climate is humid subtropical mesothermal, with hot summers, rare frosts and a tendency to rain concentration in the summer months, with no defined dry season. The average temperature in the warmer months is higher than 22 °C and lower than 18 °C in the colder months, with an average annual temperature between 20 and 21 °C (IAPAR, 1978). The pluviometric index for the trial area is higher than 1,800 mm per year. The soil classification is Rhodic Hapludox, moderate horizon A, very clayey texture, with average values of 870, 92 and 38 g kg⁻¹ of clay, silt and sand, respectively, according to Tormena et al. (2007).

The samplings were carried out in two contiguous areas: one in no-tillage (NT), and the other in a conventional tillage (CT) system, tilled with a disk plough and harrowing at a depth of 0.25 m. The NT area had been used for the previous five years for a crop rotation system with soybean, black oat, field pea and corn. The seeder used in NT was equipped with flat disks for straw chopping, fertilizer applied by a furrower and seeds with a double-disk furrower. In the area under CT, a wheat/soybean crop succession was planted. Fertilization was applied according to technical recommendations based on soil analyses. In the NT system, weeds were managed using the herbicides glifosate + chlorimuron or glifosate + flumioxazin, one week before crop planting. In CT, when infestation was high, this procedure was also applied prior to conventional tillage. After planting, weeds were controlled by herbicides applied after crop emergence, using a mixture of 500 mL fomesafen + 500 mL imazethapyr + 60 g oxasulfuron per hectare, in commercial formulations, both in NT and CT.

Initially, the trial area was surrounded with stakes and a 21 × 65 m rectangle was outlined. The central points were aligned and marked with tape, in order to obtain a regular grid of 3 × 5 m, except in two central columns where the spacing was 3 × 2.5 m, dully referenced as x and y coordinates. The total area per management system was 1,365 m², 21 m on the y and 65 m on the x axis, with a total of 128 sampling points per system. Data collection was conducted one month after the 2003/2004 soybean harvest. Undisturbed soil samples were collected by a sampler and rings with a volume of approximately 100 cm³, from the center of the 0.00-0.15 m layer. Samplings were carried out on April 30, 2004 for NT and May 09, 2004 for CT, at the cross point of the x,y coordinates.

In order to determine the soil physical properties, undisturbed samples were saturated for a period of 48 h in a tray with water to a height of ½ of the ring. After saturation, the samples were drained at -60 hPa soil water potential, using a tension table adapted from Kiehl (1979). To drain the soil samples at -100hPa tension, an extraction membrane from Richards, as described by Klute (1986) was used. Next, the samples were dried in a forced air oven for 48 h at a temperature of 110 °C. The physical properties quantified in the undisturbed samples were soil bulk density (Bd), total porosity (Tp), macropore volume (Mac) defined as pores drained at -60 hPa and micropore volume (Mic), defined as water content at -60 hPa. The following indicators were also estimated: aeration capacity of soil matrix (CAMS), an indicator of soil aeration as described by Reynolds et al. (2002), which is defined as the volume of pores drained between saturation and -100 hPa tension; soil water content at 100 hPa was used to estimate field capacity (CC), according to Reichardt (1987). Soil bulk density was determined according to Blake and Hartge (1986).

Weeds were collected using a wooden square with an area of 1.0 m². The quantified variables were: total plant shoot biomass (biomass) present in the sampled area; weed density (Dpe) or the total number of plants per m²; density of Commelina benghalensis (commelina) and density of Bidens pilosa (bidens), determining the number of these plants per m². In order to determine biomass, weeds were cut close to the ground, packaged in paper bags and dried in a forced ventilation oven at 65 °C to stable dry mass.

The measured variable data were subjected to exploratory analysis from which the mean and median were obtained as the central tendency measurements and the variance, standard deviation, and coefficient of variation as dispersion measurements. The total range and quartiles were evaluated to analyze the data distribution, and the skewness and kurtosis coefficients to verify data positioning in relation to normal distribution. These analyses were conducted with the Statistic software (Statsoft, 2000). The data were subjected to the Kolmogorov-Smirnov test for finding the approximation to normal distribution, according to Campos (1983).

Some of the variables presented a deterministic component or dependence in relation to the x or y directions, or both, which was identified by linear regression analysis using statistical software SAS (SAS, 1999). The deterministic component was removed by subtracting the value obtained with the adjusted linear regression model from the measured value, resulting in trend-free residuals, according to Vieira (2000) and Gonçalves et al. (2001). The resulting residuals of these variables were analyzed to evaluate their normal distribution and those which did not...
approximate this distribution were transformed, using
the natural logarithm function. An outlier candidate
analysis was also performed in which only the values
which effectively affected distribution, distancing
it from normal distribution, and which presented a
spatial position different from the other neighboring
values were considered discrepant, according to
Libardi et al. (1996). The data found to be discrepant
were excluded from the variographic analysis and from
cross validation, although maintained in the kriging
interpolation process.

The spatial analysis consisted in the construction
and interpretation of the cross variograms and in
the adjustment of mathematical models to explain
the combined variance structure of the data. Variographic analysis was performed with software
VarioWin 2.2 (Pannatier, 1996).

In this study, it was assumed that the data
fulfill second order stationarity which occurs when
the expectation (E) of Z(s) is constant, μ(s) = μ
and the variance (VAR) of the increment between Z(s)
and Z(s + h) is finite and independent of the spatial
position, depending only on the separation vector h,
as shown in equation 1.

\[
\text{VAR} \{Z(s_i) - Z(s_i + h)\} = \text{E}[Z(s_i) - Z(s_i + h)]^2 = 2\gamma(h)
\]  (1)

for any s_i (sample point) within the area of influence
(Gonçalves, 1997; Vieira, 2000). The crossed
covariance is defined by equation 2;

\[
C_{ij}(h) = \text{E}[Z_i(s) - \mu_i][Z_j(s + h) - \mu_j]
\]  (2)

While the spatial correlation can be described
using equation 3;

\[
\rho_{ij}(h) = C_{ij}(h)/\sqrt{C_{ii}(0) \times C_{jj}(0)}^{1/2}
\]  (3)

Cross semivariance, defined using equation 4;

\[
2\gamma_{ij}(h) = 2C_{ij}(0) - C_{ij}(h) - C_{ji}(h) = 2C_{ij}(0) - C_{ij}(-h) - C_{ji}(-h).
\]  (4)

with i and j being the two spatially dependent
variables (Webster, 1985).

The spatial correlation estimate between two
variables was performed with all possible combinations
according to Holmes (2005), by constructing the cross
variogram expressed by equation 5.

\[
N(h) \gamma_{12}^*(h) = 1/2N(h) \sum_{i=1}^{N(h)} \{Z_1(s_i) - Z_2(s_i + h)\} \{Z_2(s_i) - Z_2(s_i + h)\},
\]  (5)

in which N(h) is the number of measured value pairs
Z_i(s_i), Z_i(s_i+h) and Z_2(s_i), Z_2(s_i+h) for the variables
Z_1 and Z_2, separated by a distance corresponding
to the module of separation vector h (Vieira, 2000;
Walter et al., 2002).

The authorized mathematical functions,
according to McBratney and Webster (1986), which
were used in this study, are described in equations
6, 7, and 8:

- **Spherical model**
  \[
  \gamma(h) = C_0 + C \left(\frac{3h}{2a} \frac{1}{2} h/a\right)^3, \quad 0 < h < a
  \]  (6)

- **Exponential model**
  \[
  \gamma(h) = C_0 + C, \quad h > a
  \]  (7)

- **Gaussian model**
  \[
  \gamma(h) = C_0 + C \left(1 - \text{Exp}\left(\frac{-3h}{a}\right)\right)^2, \quad h > 0
  \]  (8)

In these equations, the coefficients of the model
adjusted to the variogram are: C_0 - Nugget effect,
which is the value of γ(h) when h = 0; C_0 + C = Sill
is the value of γ(h) when the variogram stabilizes;
C = structural variance or the difference between
the nugget effect and the sill and a = Range is the
distance to the point where the variogram
reaches the sill and represents the range of spatial
dependency, distance from which the pairs of values
taken by the variable are independent (Gonçalves,
1997; Grego and Vieira, 2005).

When one of the variables is sampled at a lower
intensity than another, co-kriging may be used to
estimate the values of non-sampled locations, from
the coefficients of the models adjusted to the direct
and cross variograms, according to Vauclin et al.
(1983) and Trangmar et al. (1985). The fitting of the
cross variogram between the variables, regardless of
order, will generate the same structure of combined
variation in space (Trangmar et al., 1985; Webster,
1985). This way, when the cross variogram is
presented between two variables, the discussion is
valid for both variables involved, regardless of the
order used to construct the cross variogram.

**RESULTS AND DISCUSSION**

The values concerning the descriptive statistics
of the soil properties evaluated for the two studied
management systems are presented in table 1. Figures 1 and 2 show the cross variograms for
all directions and respective fitted models for the
soil physical and weed properties in NT and CT,
respectively. Fourteen cross variograms were fitted
in NT and nine cross variograms in CT, in a total of 24
possible combinations involving the paired variables.

Table 2 shows the spatial correlations established
and their directions using the cross variograms
in the two soil management systems. A negative
spatial correlation was verified between the total
Dpe (commelina and bidens) with Bd, and no
correlation between biomass and Bd was found
in NT. The direction of spatial correlation shows that, as Bd increases, there is a decrease in Dpe. Therefore, while weeds have a higher capability of adapting to adverse environments, in this soil management system they behave similarly to crop plants, with reduced growth and development in more compacted soil.

In CT, the Dpe correlated positively with Bd (Figure 2). These results reveal the effects of soil management on these relationships, as in NT there was no correlation between biomass and Bd and the correlation between Dpe and Bd was negative, while in CT the correlation was positive. A change in the direction of the correlation was verified: in CT, the increase in Bd promotes an increase in species diversity and a reduction in total biomass. This behavior may evidence the existence of tenuous correlations that will not persist over time, as discussed by Walter et al. (2002), because soil tillage constantly modifies the soil structure. This induces different correlations, even within the same soil, as stated by Walter et al. (2002) and Maestre et al. (2005), but influenced by the management system. These results agree with those of Walter et al. (2002) who found a negative correlation between the density of Viola arvensis and soil clay contents, although Heisel et al. (1999) detected no significant correlation between total Dpe and soil properties.

All weed variables were spatially correlated with Tp in the NT system (Figure 1). A negative correlation was verified between biomass and Tp, and a positive one between Dpe and the density of commelina and bidens with Tp. These results suggest that the weeds take advantage of conditions of higher soil porosity, establishing a larger diversity of species. The positive correlation between Dpe and Pt, in spite of maintaining a lower total biomass, is an indicator of this relationship in NT. The larger diversity is an indicator of larger environmental stability, as suggested by Odum (1988).

In CT, the weed biomass correlated negatively with Tp, with a variogram range of 42 m. The direction of the correlation was similar to the NT system, however, with a wider range in CT, pointing to a larger spatial dependency in this system. While there was no significant correlation between Tp and Dpe in this management system, it can be said that there is a decrease in total biomass with an increase in Tp (Table 1). The higher Tp in NT showed a larger species diversity, indicated by the positive correlation between Tp and Dpe (Figure 1). This effect was not repeated in CT, possibly due to the lower structural stability of the soil in this system, with lower soil porosity, as well as to the physical effect of periodical soil tillage. Heisel et al. (1999) and Walter et al. (2002) presented different results for correlations of the same studied variables, evidencing their specificity.

The Dpe variables correlated positively with macroporosity (Mac) in the NT system (Figure 1). The larger range of spatial dependency between bidens and Mac, of 20.5 m, may be related with the dispersal mechanism of this species, which allows the seeds to be taken across larger distances when compared to commelina, with a range of 10.5 m. The larger quantity of macropores in the soil indicates
Figure 1. Cross variograms and mathematical models adjusted to weed variables and physical properties of the soil in no-tillage system. Bd: soil bulk density (kg dm\(^{-3}\)); Tp: total porosity (m\(^3\) m\(^{-3}\)); Mac: macropores equivalent to the pores drained at -60 hPa tension (m\(^3\) m\(^{-3}\)); Micro: microporosity (m\(^3\) m\(^{-3}\)); CAMS: pores drained at -100 hPa tension (m\(^3\) m\(^{-3}\)); CC: estimated water content at -100 hPa tension (m\(^3\) m\(^{-3}\)); Dpe: weed density (plants m\(^{-2}\)); biomass: dry matter of weed shoots (g m\(^{-2}\)); commelina: density of *Commelina benghalensis* plants (plants m\(^{-2}\)); bidens: density of *Bidens pilosa* plants (plants m\(^{-2}\)).
that good drainage after soil saturation, allowing a faster diffusion of oxygen to the roots of the weeds, according to Tormena et al. (1999a,b). The positive correlation between total density and plant species with Tp and Mac indicates that this soil condition is favorable for plant development, which is used by weeds in their establishment and development.

The biomass was negatively correlated with Mac and positively with Dpe in CT (Figure 2). As Mac increases, there is a positive effect on the increase in diversity of weed species, which is a good indicator of the greater stability of agro-ecosystems under no-tillage. This decrease in total biomass is possibly due to the interspecific competition between

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**Figure 2.** Cross variograms and mathematical models adjusted to weed variables and soil physical properties in conventional tillage. Bd: soil bulk density (kg dm⁻³); Tp: total porosity (m³ m⁻³); Mac: macroporosity or equivalent of pores drained at -60 hPa tension (m³ m⁻³); Mic: microporosity (m³ m⁻³); CAMS: pores drained at -100 hPa tension (m³ m⁻³); CC: estimated water content at -100 hPa tension (m³ m⁻³); Dpe: weed density (plants m⁻²); biomass: dry matter of weed shoots (g m⁻²).
the shoot and the root system, as established by Rizzardi et al. (2001). The variable biomass of weeds correlated positively with CAMS and no correlations were verified between these and the other variables in the NT system, while CAMS was negatively correlated with Dpe in CT. The variable Dpe correlated negatively with CC in CT. There is a contradiction between the behaviors of these two variables in relation to CAMS, since when CAMS increases the water content at field capacity decreases. Therefore, the expected direction of the correlation would be the inverse between the Dpe and CAMS and CC. In the NT system, there was no significant correlation between these variables, and it can be inferred that these correlations in CT are weak correlations, which will not persist over time, as pointed out by Walter et al. (2002).

In the NT system, commelina density correlated negatively with the water content at field capacity (Figure 1). Analyzing the positive correlations of this species with Tp and Mac and the negative correlation with Bd and water content at field capacity, it can be inferred that this plant has a better development in well drained, uncompacted soils, in agreement with its dispersal system, with airborne and subterranean seeds and rhyzomes, as described by Walker and Evenson (1985a,b). Although crop plants develop better under the same conditions and can hamper the development of this species, it is able to establish during the crop cycle even under shaded conditions and starts growing abundantly immediately after harvest (Schaffrath and Miller, 2000).

A negative and positive correlation was verified between biomass and Dpe, respectively, with microporosity (Mic) the NT system. The negative correlation of weed biomass with Tp and soil Mic, as shown in table 1, may indicate a close relationship between weed diversity and biomass production of these plants with these variables. Generally, it can be said that, based on the correlations observed between Tp and weed variables, increases in Tp result in increases in Dpe and in the density of the species studied, although biomass decreases. This effect, expanding the number of species present but reducing the total biomass, may be the result of an interspecific competition between weeds, i.e., species diversity is larger but with a lower production of total biomass, as suggested by Maestre et al. (2005). It can also be inferred that the effect of the increase in Tp is beneficial in agronomic terms, as a lower weed biomass production may indicate a decrease in competition between weeds and crop plants (Rizzardi et al., 2001).

A larger diversity of species is an indicator of greater environmental stability, a crucial condition for establishing agro-ecosystems managed sustainably. In CT, Mic was positively correlated with biomass and negatively with Dpe. These correlations occur in the reverse direction than the correlation of Pt with biomass and Mac with biomass. These relationships indicate that correlations in CT tend to be unstable over time, or short term correlations, requiring, for that reason, a longer evaluation period to verify the persistence of these correlations, as suggested by Walter et al. (2002).

### CONCLUSIONS

About half the variables analyzed were spatially correlated, mainly in the NT system. In these cases, cokriging may be used to map one variable from another.

The soil management system influenced the spatial correlations.

A larger diversity of weeds was observed in the no-tillage system, and a larger weed biomass in conventional tillage.

The spatially correlated variables can be used to draw conclusions from one another in the studied management systems.

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