Spatial analysis of hydraulic conductivity measured using disc infiltrometers

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Abstract
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Disciplines
Agriculture | Bioresource and Agricultural Engineering | Hydrology | Water Resource Management

Comments
Spatial analysis of hydraulic conductivity measured using disc infiltrometers

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Abstract. Spatial variability of surface hydraulic properties and the extrinsic (e.g., traffic, cropping, etc.) and intrinsic (e.g., soil type, pore size distribution, etc.) factors associated with these properties are important for infiltration and runoff processes in agricultural fields. Disc infiltrometers measured infiltration at 296 sites arranged on two parallel transects. To examine and differentiate the factors contributing to spatial structure under different field conditions these measurements were made in the corn rows, no-track interrows, and wheel track interrows of the field using four different soil water tensions Ψ (0, 30, 60, and 150 mm). Unsaturated hydraulic conductivity (K) and saturated hydraulic conductivity (Ks) were maximum in the corn rows and minimum in wheel track interrows, with no-track interrows intermediate. Exponents (α parameters) of Ks and K relationships (K = Ks exp -αΨ) for corn rows and no-track interrows were not significantly different from each other but were significantly different from α for the wheel track interrows at P = 0.01 level. Spatial variability of K and Ks values showed some pseudoproportional effect in nugget variance for all three field conditions. No-track interrows clearly showed an inverse trend for semivariogram of K with changing tension (Ψ) values, whereas differences were found for corn rows and wheel traffic interrows. The spatial structure of α for all three field conditions were mostly white noise. Under corn rows, in addition to random variation, a small five-row periodic variation at the P = 0.20 level, matching the five-row traffic configuration, was discovered. The spatial structure of α was influenced by soil type for the no-track interrows. Spatial structure was absent in wheel track interrows, indicating the destruction of pore structure due to compaction.

Introduction

Surface conditions of a field influence the infiltration of water into the soil profile. At the beginning and receding periods of a storm event, when the soil is not saturated, unsaturated hydraulic conductivity (K) of the thin surface soil affects the vertical infiltration process. Moreover, saturated hydraulic conductivity (Ks) of the surface layer will influence overland runoff produced relative to the downward infiltration in the subsoil during a storm event. Earlier studies [Edwards et al., 1979, 1988; Dick et al., 1989; Blevins et al., 1990] have shown that infiltration of water into soil is directly related to soil macroporosity. These studies indicated that runoff is much less when soil contains a large number of macropores, compared with soil that has few macropores.

Macropores are important for root growth and solute and water movement [Beven and Germann, 1982]. Studies to quantify macropore flow under rain-fed conditions [Watson and Luxmoore, 1986; Wilson and Luxmoore, 1988] revealed that more than 70% of water flux can move through the macropores. Meek et al. [1989] found 2 to 3 times increase in the infiltration rates over a 3-year period for a field under flood irrigation practice planted to alfalfa. The authors attributed the increase to flow through root channels. Shir-mohammadi and Skaggs [1984] conducted experiments to determine the effects of different factors including plant cover on infiltration into columns of fine sand. They found that rescuing roots loosened the soil and increased the hydraulic conductivity by an average of 40% over soybeans and 80% over bare soil. Surface soil conditions, resulting from different tillage practices and/or wheel traffic compaction, likely will contribute to spatial variation of the soil macroporosity, thereby altering the hydraulic properties. Disregarding these factors may confound the interpretation of infiltration and runoff results.

Barley [1954] found that the growth of corn roots compacted the soil between the root channels and greatly reduced the permeability, and macropores formed in corn rows as a result of root decomposition increased the permeability. Meek et al. [1990] showed that the depth and size of the alfalfa taproot system produced an extensive macropore flow system, resulting in high infiltration rates when the roots decomposed. Warner and Young [1991] concluded that corn roots provided preferential pathways for water movement through the topsoil. Wetting and drying cycles [Akram and Kemper, 1979], freezing and thawing [Akram and Kemper, 1979; Carter, 1988], and earthworm activity [Trojan and Linden, 1992] may also contribute to macropore generation,
pore structure manipulation, and increased infiltration rate. Additionally, wheel traffic by agricultural machinery may contribute to changes in infiltration capacity of soil. Wager and Denton [1989] found an 86% reduction in $K_s$ in tracked areas compared with that in untracked interrow areas.

On a spatial scale the difference in infiltration between row and interrow areas might be influenced by mutually exclusive intrinsic/extrinsic factors such as soil type, tillage practice, root growth/decay, worm holes, and wheel traffic. Mohanty et al. [1994] found the difference in infiltration because of measurement methods used. Under no-till conditions, macropore networks can be well developed (thereby short-circuiting the vertical movement of water) whereas macropores are poorly developed in conventionally tilled soil [Ehlers, 1975; Edwards et al., 1979, 1988; Germann et al., 1984; Zachmann et al., 1987; West et al., 1991]. Edwards et al. [1979, 1988] showed that the number of macropores per unit soil area, as well as the diameter and depth of these pores, greatly influence infiltration and surface runoff. Therefore transport of water in an agricultural field can be viewed as a spatiotemporal process. Identifying changes in the hydraulic conductivity among row, interrow, and wheel traffic conditions on a spatial scale will help unmask some of the intrinsic factors (e.g., pore size distribution and soil type) and extrinsic factors (e.g., root growth and decay and traffic condition) contributing to spatial variability in the infiltration process. Knowledge of these factors will help hydrologists and soil physicists to model infiltration and overland flow processes more accurately.

Although some studies have investigated the effect of traffic and crop rows on infiltration and solute transport properties of soil [Meek et al., 1989, 1990, 1992a, b; Timlin et al., 1992; van Wesenbeeck and Kachanoski, 1988], to our knowledge no study has been conducted to investigate the effect of these factors (crop row, interrow, and wheel traffic) on spatial variability in soil hydraulic properties ($K$, $K_s$, and $\alpha$). The relation between hydraulic conductivity at different tensions and saturation at a particular site is often described using $K = K_s \exp^{-\alpha T}$. Thus in addition to $K_s$ and $K$ the exponent in the relationship, $\alpha$, is also a site-specific hydraulic parameter of soil and may be subjected to spatial variability [Russo and Bouton, 1992, and references therein].

This study was conducted in a glacial till soil in central Iowa. Natural processes involved in the weathering of a glacial till soil are highly variable and complex [Terszaghi and Peck, 1967], but man-made disturbances may make the infiltration capacities of these soils either more or less heterogenous and/or correlated, spatially. Wilson and Luxmoore [1988] suggested that the variability of infiltration rates did not decrease at higher tensions, and they concluded that smaller pores are as variable as larger pores. This result is inconsistent with the conclusions of Clothier and White [1981]. Considering these contradictory results, Ankeny et al. [1990] pointed out that without knowledge of both measurement variability and site variability, the cause of variability in infiltration rates among pore-size classes cannot be determined. Therefore this study was conducted considering two primary objectives: (1) to compare the hydraulic conductivity measurements ($K$ at different tensions ($\Psi$) and $K_s$) under crop row, wheel track interrow, and no-track interrow conditions in an agricultural field, and (2) to study the spatial variability in $K$, $K_s$, and $\alpha$ and investigate the underlying (intrinsic) governing factors under these three (extrinsic) surface conditions.

Field Data Collection Procedure

Micro pores, mesopores, and macropores are defined in several ways [Luxmoore, 1981]. Ankeny et al. [1990] defined macropores as pores that empty at <150 mm of water tension. The capillary-rise equation predicts that pores of 0.2 mm nominal diameter or larger will drain at 150-mm tension. This pore size was chosen as important for root growth [Hackett, 1969; Russell, 1977] and preferential solute transport [Scotter, 1978]. The development of the tension infiltrometer has enabled researchers to quantify the effective macroporosity and equivalent cylindrical diameter macropores [Clothier and White, 1981; Ankeny et al., 1988; Perron and White, 1988]. Controlling tension at the soil surface by a tension infiltrometer limits the size of pores that are active in conducting water [Clothier and White, 1981]. Sequential increase of tension leads to incremental draining of smaller and smaller pores. Infiltration rate decreases as more water-conducting pores are emptied.

In this study, infiltration rates at selected soil water tensions (0, 30, 60, 150 mm) were measured with automated disc infiltrometers [Ankeny et al., 1988; Ankeny, 1992; Priekaitis et al., 1992]. Ponded and tension infiltrometers with a 76.2-mm-diameter base were used. Measurements were made along two parallel transects running in a SE-NW direction at the Agronomy and Agricultural Engineering Research Center near Boone in central Iowa (Figure 1). Corn rows and interrows were measured on two parallel transects 1 m apart (Figures 1 and 2). Figure 2 shows the five-row configuration of tractor and equipment traffic. The distance between consecutive rows and consecutive interrows was 76 cm. Infiltration measurements were made at the center of all corn rows (irrespective of old corn plant locations) and interrows. We chose transects across rows rather than along rows to avoid any localized treatment effects (e.g., traffic, chemical management, soil type with varying bulk density and thus differently compacted by wheel traffic, any misdriven traffic interrow, and compaction because of different weights of tractors and equipments used for different chemical management practices) on $K$. All the measurements were limited to the no-tillage management practice to avoid any tillage effect on the spatial variability of $K$, $K_s$, and $\alpha$. No-tillage and no-post plant cultivation were practiced for 8 years in this field prior to this experiment. Field plots were established for row crop production with continuous corn during this 8-year period. Soil type in this field is predominantly silt loam that belongs to the Niccollet-Clarion-Webster soil series. Maximum land slope of the area where transects were located was 2%. This study was conducted in May 1990 during the corn planting season. Infiltration measurements at 296 sites were completed in less than a week, minimizing the temporal variability in the data. To accomplish these measurements in such a short period of time, 18 tension infiltrometers and six ponded infiltrometers were used simultaneously. All these disc infiltrometers were calibrated with a reference manometer (as described by Ankeny [1992]) and used randomly within two transects.

Disc infiltrometer readings were taken at 160 corn row sites and 136 interrow sites at 0, 30, 60, and 150 mm tensions at each site. At each infiltration measurement site an area
approximately 25 to 30 cm in diameter was cleared with a hand trowel to a depth of 20 to 30 mm and leveled. This site preparation method minimizes compaction and smearing due to initial dry conditions. Two layers of cheesecloth (of approximately 2 mm pore diameter) were placed on the soil surface before wetting, to minimize slaking of soil into the macropores. A ponded infiltrometer was set, and saturated (0-mm tension) flow measurements were taken followed by tension infiltrometer measurements from low to high tension (30 to 150 mm) on a 7.62-cm-diameter circle on the cleared soil surface. A detailed ponded/tension infiltrometer measurement procedure is outlined by Ankeny [1992]. A wet to dry sequence of measurements was followed to reduce the antecedent water potential effects at low infiltration rates. By conducting saturated measurements first, the wetting front advances as rapidly as possible, and the assumption of a unit gradient below the device is most valid. Unsaturated hydraulic conductivity \( K \) at different tensions was estimated from the corresponding infiltration rate by using the relationship developed by Ankeny et al. [1991]. Then \( \alpha \) values for (three) different tensions were calculated from the \( K_s \) and \( K \) relationships. An average \( \alpha \) was computed using \(\alpha_{30}, \alpha_{60}, \text{and} \alpha_{150}. \) This multiple-tension-based measurement allowed evaluation of treatment effects on different pore sizes. After infiltration measurements were completed for all four tensions at one site, a 7.62-cm soil core was taken from the (infiltration) measurement site and visually examined for root growth and macropores. In corn row locations, 149, 153, 146, and 138 data points could be gathered for analysis at 0-, 30-, 60-, and 150-mm tensions. The reason for missing values are human error and/or instrument failure. Similarly, 54, 72, 72, and 59 data points were available for no-track interrows and 27, 35, 33, and 17 for wheel track interrows.

**Theory of Spatial Analysis**

Geostatistics offer a variety of techniques to fabricate models for transport processes as realizations of space-time random functions. Two basic assumptions used in geostatistics are ergodicity and stationarity of the data. Ergodicity refers to the assumption that the unique realization available behaves in space with the same probability density function.
as the ensemble of possible realizations. Stationarity means that any statistical property of the medium does not change with translation thus it is stationary in space, that is, the property is said to be statistically homogenous. Because second-order stationarity [Journel and Huijbregts, 1978, p. 32] has often been violated [Horowitz and Hillel, 1983], a weaker intrinsic hypothesis [Matheron, 1963] is often used to simplify the analysis. Under this assumption the classical semivariogram estimator [Matheron, 1963] is defined as

\[ \gamma(h) = \frac{1}{2N(h)} \sum_{i=1}^{N(h)} [Z(x_i) - Z(x_i + h)]^2, \] (1)

where

- \( \gamma(h) \) semivariogram estimator for lag distance class \( h \);
- \( Z(x_i) \) measured sample value at point \( x_i \);
- \( Z(x_i + h) \) measured sample value at point \( x_i + h \);
- \( N(h) \) total number of sample couples for the interval \( h \).

Different theoretical models such as linear, spherical, and exponential are most commonly used in the field of soil science and hydrology to describe the semivariogram function of regionalized variables. In real-life situations, spatial behavior of soil properties is often so complex that a simple theoretical model could hardly suffice in fitting the data. Mohanty et al. [1991] and Oliver and Webster [1986] used nested models (\( \gamma(h) \)) to represent various kinds of field situations. This structure can be defined as

\[ \gamma(h) = \gamma_1(h) + \gamma_2(h) + \gamma_3(h) + \cdots + \gamma_n(h), \] (2)

where \( \gamma_1(h), \gamma_2(h), \cdots, \gamma_n(h) \) are contributions from several semivariogram models. Small-scale variability (e.g., microheterogeneity), large-scale variability (e.g., soil type effect), directional trend (e.g., due to land slope), and periodicity (e.g., due to field traffic configuration) are some of the typical variations found in hydrology and soil science. All of these theoretical models or their tailored combinations, when best fit to the estimated sample semivariogram, reveal the intricacy of spatial pattern of the soil property. Contributing factors of these spatial structures could be investigated by studying the history and geography of the site. Besides revealing spatial structure of a series of observations, a semivariogram is an appropriate tool to estimate unbiased values at unrecorded points using kriging procedures and to account for possible drifts in the property of interest.

In addition to semivariogram, Gajem et al. [1981], Sisson and Wierenga [1981], Russo and Bresler [1981], and Vauclin et al. [1982] considered autocorrelation functions to describe the spatial structure of several soil properties. The autocorrelation function is defined as

\[ \rho(h) = \frac{\text{Cov}(Z(x_i), Z(x_i + h))}{\sigma^2}. \] (3)

Generally, the joint distribution of \( Z(x_i) \) and \( Z(x_i + h) \) is not known, and \( \rho(h) \) will have to be estimated from the available set of data. In that sample, autocorrelation function \( r(k) \) can be estimated by different estimators defined by Davis [1973], Cliff and Ord [1973], and Haan [1977]. Autocorrelation function [Davis, 1973] is a more traditional approach to autocorrelation and defined as

\[
\begin{align*}
    r(k) &= \frac{\sum_{i=1}^{n-k} Z(x_i) Z(x_i + k) - \sum_{i=1}^{n-k} Z(x_i) \sum_{i=1}^{n-k} Z(x_i + k)}{\sqrt{\sum_{i=1}^{n} Z(x_i)^2 - \sum_{i=1}^{n-k} Z(x_i) \sum_{i=1}^{n-k} Z(x_i + k)}} \frac{1}{n(n-1)} \quad (4)
\end{align*}
\]

where

- \( r(k) \) is the value of the autocorrelation function at lag \( k \) such that \( h = k \Delta x \) for observations spaced at regular intervals \( \Delta x \) along a transect of \( n \) observations. \( Z(x_i) \) and \( Z(x_{i+k}) \) are the values of the observations at the \( i \)th and \( i + k \)th position, respectively. Although it requires an assumption of second-order stationarity, the autocorrelation function has certain advantages over other spatial analysis methods. In particular, its values are normalized to the range \([-1, 1]\), making it easier to interpret the values and to compare several replications. The confidence limits on \( \rho(h) \) at probability level \( P \) are then estimated, as defined by Haan [1977]:

- Lower limit \([-1 - t_{1-P/2} \sqrt{n-2}/(n-1)]\)
- Upper limit \([-1 + t_{1-P/2} \sqrt{n-2}/(n-1)]\)

If the calculated \( r(k) \) falls outside these confidence limits, the hypothesis that \( \rho(h) \) is zero \((H_0: \rho(h) = 0 \text{ versus } H_1: \rho(h) \neq 0)\) is rejected at probability level \( P \).

Webster [1977] and Vauclin et al. [1982] used spectral analysis to determine the periodicity in data by considering the spectral density function (Fourier transform of the autocorrelation function) to describe variations along a transect. Spectral density function \( (S(f)) \) is defined for a discrete spatial series as

\[ S(f) = \Delta x \left[ r(0) + 2 \sum_{k=1}^{m} r(k) \cos(2nkf\Delta x) \right], \] (7)

where \( m \) is the maximum number of correlation lags. Values of \( S(f) \) should be computed only for frequencies \( f \) given by

\[ f = \frac{k}{m} f_N, \] (8)

where \( f_N (= 1/2\Delta x) \) is the Nyquist frequency. The estimates of \( S(f) \), however, need to be smoothed by filtering techniques before extensive interpretation.
In our analysis all the above methods were used in different instances to supplement each other for interpreting the result. Note that all these spatial analyses are preceded by a Hawkins statistics operation in order to remove any existing outlier(s) in any data set. Also, possible drift (first-order increments of $Z$) was examined by experimentally computing $E[Z(x) - Z(x + h)]$, as suggested by David [1977], prior to semivariogram and autocorrelation computation.

Results and Discussion

A classical analysis of variance (ANOVA) (Table 1) was made for $K$ at different tensions and $\alpha$ to determine whether there is significant treatment difference between three field conditions: corn row, no-track interrow, and wheel track interrow. In all three cases, results indicate significant between-treatment differences at $P = 0.01$. Following this finding, all further spatial analyses were performed for individual treatments (field conditions) to eliminate any intermixing of within variability and between variability of the treatments.

Table 1. Analysis of Variance of $K$ at 0-, 30-, 60-, and 150-mm Soil Water Tension and of $\alpha$

<table>
<thead>
<tr>
<th>Tension, mm</th>
<th>Between-Treatment Variation</th>
<th>Within-Treatment Variation</th>
<th>Total Variation</th>
<th>Between-Treatment Mean Square</th>
<th>Within-Treatment Mean Square</th>
<th>$F^{**}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$K$</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0</td>
<td>200265 2</td>
<td>1893167 227</td>
<td>2093432 229</td>
<td>100132</td>
<td>8340</td>
<td>12.0</td>
</tr>
<tr>
<td>30</td>
<td>915 2</td>
<td>4207 257</td>
<td>5122 259</td>
<td>407.7</td>
<td>16.4</td>
<td>24.9</td>
</tr>
<tr>
<td>60</td>
<td>124 2</td>
<td>754 248</td>
<td>878 250</td>
<td>58</td>
<td>3</td>
<td>20.6</td>
</tr>
<tr>
<td>150</td>
<td>11.6 2</td>
<td>97.5 211</td>
<td>109 213</td>
<td>0.013</td>
<td>0.0004</td>
<td>32.6</td>
</tr>
<tr>
<td>$\alpha$</td>
<td>0.026 2</td>
<td>0.092 227</td>
<td>0.118 229</td>
<td>0.013</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

*a Degrees of freedom, d.f.  
**Significant at $P = 0.01$ level.

In our analysis all the above methods were used in different instances to supplement each other for interpreting the result. Note that all these spatial analyses are preceded by a Hawkins statistics operation in order to remove any existing outlier(s) in any data set. Also, possible drift (first-order increments of $Z$) was examined by experimentally computing $E[Z(x) - Z(x + h)]$, as suggested by David [1977], prior to semivariogram and autocorrelation computation.

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Table 2. Summary Statistics of $K$ at Different Soil Water Tension for Corn Row, No-Track Interrow and Wheel Track Interrow

<table>
<thead>
<tr>
<th>Tension, mm</th>
<th>Corn Row</th>
<th>No-Track Interrow</th>
<th>Wheel-Track Interrow</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Moments</td>
<td>N*</td>
<td>Mean</td>
</tr>
<tr>
<td>0</td>
<td>Mean</td>
<td>110.7</td>
<td>27.2</td>
</tr>
<tr>
<td>30</td>
<td>Maximum</td>
<td>723.2</td>
<td>14.6</td>
</tr>
<tr>
<td>60</td>
<td>Minimum</td>
<td>2.8</td>
<td>0.1</td>
</tr>
<tr>
<td>150</td>
<td>Variance</td>
<td>10241.4</td>
<td>22.1</td>
</tr>
<tr>
<td>0</td>
<td>Coefficient of variation, %</td>
<td>91.4</td>
<td>67.7</td>
</tr>
<tr>
<td>30</td>
<td>W Normal†</td>
<td>0.799* 0.841* 0.862* 0.826*</td>
<td>0.866*</td>
</tr>
<tr>
<td>60</td>
<td>W Normal†</td>
<td>0.706* 0.881* 0.848* 0.866*</td>
<td>0.710*</td>
</tr>
<tr>
<td>150</td>
<td>W Normal†</td>
<td>0.706* 0.881* 0.848* 0.866*</td>
<td>0.710*</td>
</tr>
</tbody>
</table>

$k$ (hydraulic conductivity) values are in micrometers per second.

*Number of data points.
†Normality test using the Shapiro-Wilk $W$ statistic.
‡Not normally distributed at a confidence level $>0.90$. 

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<table>
<thead>
<tr>
<th>Tension, mm</th>
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<th>Corn Row</th>
<th>No-Track Interrow</th>
<th>Wheel-Track Interrow</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>Mean</td>
<td>110.7</td>
<td>27.2</td>
<td>1.0</td>
</tr>
<tr>
<td>30</td>
<td>Maximum</td>
<td>723.2</td>
<td>14.6</td>
<td>7.9</td>
</tr>
<tr>
<td>60</td>
<td>Minimum</td>
<td>2.8</td>
<td>0.1</td>
<td>0.2</td>
</tr>
<tr>
<td>150</td>
<td>Variance</td>
<td>10241.4</td>
<td>22.1</td>
<td>4.0</td>
</tr>
</tbody>
</table>

$K$ (hydraulic conductivity) values are in micrometers per second.

*Number of data points.
†Normality test using the Shapiro-Wilk $W$ statistic.
‡Not normally distributed at a confidence level $>0.90$. 

Table 3 shows statistics for the $\alpha$ parameter under the three field conditions. Mean values for $\alpha$ for corn row and no-track interrow were not significantly different at the 0.01 probability level, but both were significantly different from the $\alpha$ for the wheel track interrow. CV values, however, were similar for all three field conditions (44% for corn row, 47% for no-track interrow, and 42% for wheel track interrow). The CV of $\alpha$ was found to be much smaller than the CV of $K$ and $K_s$ for all three field conditions. This indicates that $\alpha$ is a more homoscedastic parameter. Furthermore, a larger range of $\alpha$ was observed for wheel track interrow compared to the other two field conditions. The Shapiro-Wilk normality test for $\alpha$ showed a better fit with the lognormal distribution than the normal distribution. Thus follow-up spatial analyses were conducted for log$_e(\alpha)$ in corn rows, no-track interrows, and wheel track interrows.

Spatial Analysis of $K$

Analysis of drift along the transect indicated no significant drift in $K$ at any tension under any field condition. Figure 3 shows sample semivariograms calculated for log $K$ at four different tensions (0, 30, 60, and 150 mm) under three different field conditions (corn row, no-track interrow, and wheel track interrow).
wheel track interrow). The lag distance (Figure 3 and figures hereafter) is intentionally expressed as row/interrow counts (one row/interrow count is 76 cm) to facilitate our site reexamination for possible intrinsic factor(s) contributing to variability in log K, K_s, and a. Sample semivariograms of log K at all tensions under each field condition were found to be mostly nugget with a certain amount of periodic behavior. These plots indicated pseudoproportional effects in nugget variance for all three field conditions, where proportional effect defines the relative shift in variance with respect to mean. Different trends in the semivariogram of log K with respect to tension were found for the three field conditions.

Under no-track interrow the semivariogram estimate for log K decreased with tension (i.e., maximum for K_0 and minimum for K_150). Thus variability in pore sizes decreases too. This behavior might be explained as follows. At 0-mm tension, pores of all sizes (micropores to macropores) conduct water. As the water tension is increased, the conducting pores are those with smaller diameter. At 150-mm tension, for example, pore sizes less than 0.2 mm will conduct water. Because macropores are root- or earthworm-created phenomena, their variability would be larger than that of natural (smaller) pores inherent in the soil. Under no-track interrows this trend of variability of inherited pore structure (of soil) was most prominent because of least external physical disturbance.

For 0-, 30-, and 60-mm tensions, corn rows showed a trend (with respect to tension) similar to no-track interrows. At 150-mm tension, semivariogram estimates were found greater than that for log K at 30- and 60-mm tensions. We suspect that this behavior might be the variability in pore size distribution because of the presence or absence of old corn roots. Moreover, our results support Hamlett et al. [1986], Cressie and Horton [1987], Gotway and Cressie [1990], and Mohanty and Kanwar [1994], who have reported that crop rows make some contribution to differing hydrologic/chemical properties.

The maximum semivariogram occurred at 30-mm tension in wheel track interrows. The reason for this behavior is not known. We suspect that many of the bigger pores that conduct water at a tension equal to or less than 30 mm were smeared and clogged from wheel traffic, thus reducing the semivariogram of K at 0-mm tension relative to that of K at 30-mm tension. Moreover, partial sealing of some macropores to different degrees due to differing compaction (load), which otherwise conduct at saturation, might increase their ability to conduct at 30-mm tension, causing additional variability in K at that tension. Ankeny et al. [1990] found that larger macropores transported a greater portion of the total water flow in untrafficked soil than in trafficked soil, which reinforces the concept that larger pores are more easily destroyed by wheel traffic than smaller pores. Studies by Meek et al. [1989, 1992a, b] found that macropores or biopores allowed water flow to bypass soil compacted by harvest traffic, reducing hydraulic conductivity of the surface layers by 60% compared to no-harvest traffic treatments. From all these findings, including ours, we suggest that compaction destroys a larger percentage of pores which carry water at lower tension (larger pores) than pores which carry water at higher tensions (smaller pores).

Spatial Analysis of a

No apparent drift in a was observed along the transect, when examined for corn row, no-track interrow, and wheel track interrow conditions. Sample semivariogram of a were estimated (Figures 4, 5, and 6) for corn rows, no-track interrows, and wheel track interrows. For all field conditions these semivariograms (up to half of the transect length) showed nugget behavior. Correlograms (autocorrelation function versus lag) up to a quarter of the transect length (Figures 4, 5, and 6) confirmed this finding at the 95% confidence level with few minor deviations.

Some of the salient features or the correlogram under corn row condition are as follows. Except for lag 1, no other r(k)

![Figure 3](image-url)
was found significant at the 95% confidence level. A steep initial decay of $r(k)$ followed by a more gradual decay up to lag 5 indicates a large random variation with a small structural variation with a range equal to five row counts. Beyond lag 5, $r(k)$ was found to be somewhat positive periodic up to a lag of 25 row counts. This periodic behavior gradually damped out beyond this lag with some fluctuations around $r(k) = 0$ contained within the 95% confidence limit. To investigate the significance of this periodic variation, several sets of confidence limits around $r(k)$ were computed using (5) and (6). Figure 7a indicates that a periodic variation (of five-row count wavelength) is significant at the 80% confidence level. This periodic variation was reconfirmed by a follow-up spectral analysis (Figure 7b). Spectral density function showed peak concentration of variance at the frequency $f = 0.2$ row count$^{-1}$, indicating a periodic structure at five (equal to $1/0.2$) row counts. The possible reason to induce this periodic behavior that we could suggest is the five-row traffic configuration (in Figure 2). Moreover, to evaluate the contribution of nugget versus the signal (structure) in the total variance of $\log \alpha$ under corn row condition, a first-order autoregressive model plus a white noise was fitted to $r(k)$ at short lag distance (Figure 7). The best fit exponential model is

$$
\rho = a \exp \left( -\frac{|h|}{\lambda} \right)
$$

when $|h|$ is the absolute value of lag distance, equal to $k\Delta x$, $a = 0.33$, and $\lambda = 3$. Thus the signal to white noise ratio calculated by $a/(1 - a) = 0.5$ indicates the dominance of random variation over the structural variation.

For no-track interrow, random variation was observed in the semivariogram and correlogram of $\log \alpha$. However, close examination of the semivariogram and correlogram showed two distinctly different populations (clusters) of sample semivariograms for the lags below 40 and above 40. The white noise became severely erratic beyond lag 40. Researchers usually avoid interpreting this type of unusual phenomena, as it may confound any artifact in the interpretation. However, we wanted to probe the intricacy of this pattern by reexamining the experimental site for possible intrinsic factors causing this behavior. Overlaying the soil map of the site on top of the infiltration measurement sites (transect) map showed that part of the transect lies in Nicollet soil and part lies in Clarion soil (Figure 1). While Clarion soils are well drained in Nicollet-Clarion-Webster series, Nicollet soils are somewhat poorly drained. Studies by Campbell [1978] suggest that soil type may be a possible
influential factor for $K$ distribution in the soil profile. To examine any soil type effect on spatial structure of $\alpha$ we divided the transects for Nicollet and Clarion soils. Mean and variance of $\alpha$ for both soils under different field conditions are presented in Table 4. Figure 8 shows the resulting semivariogram of log $\alpha$ for two soils under corn row and no-track interrow conditions. Under wheel track interrow we did not conduct soil type based semivariogram analysis because of too small sample size. Semivariograms for both soils showed random variation under corn row and no-track interrow conditions. Evidently, under corn row condition, semivariograms for two soils merely indicate a small proportional effect. On the other hand, in no-track interrows, Clarion soil showed more sharp peaks and drops in white noise in contrast to smoother variation in Nicollet soil. From this finding we suggest that these two different soil types contribute to the dual regional behavior of the semivariogram in Figure 5. Unlike corn rows, where the semivariogram shows a periodic effect due to five-row traffic configuration and other unknown intrinsic reasons, the semivariogram for no-track interrows shows the contribution of soil type (Figure 8). The existence of a soil-type effect in no-track interrows indicates minimum extrinsic factors contributing to the spatial variability, as variography is a soil-inherited phenomenon. Based on this finding, one would expect a soil-type effect in the semivariograms for corn rows and wheel track interrows. But, a soil-type effect was diluted or completely removed because of other dominant external factors (e.g., presence of macropores and roots in corn rows and compaction and destruction of pore structure in wheel track interrows).

For the wheel track interrow, no structure was found in the sample semivariogram and correlogram (Figure 6). Moreover, lack of enough pairs at different lags made the sample semivariogram and autocorrelation more unreliable.

Conclusions

A series of infiltration measurements at 0-, 30-, 60-, and 150-mm water tensions was made on the soil surface at 296 sites arranged on two parallel transects in a glacial till soil in

| Table 4. Statistics of $\alpha$ Under Nicollet and Clarion Soils |
|---------------------|---------------------|---------------------|
| Treatment          | Statistics            | Nicollet            | Clarion            |
| Corn Row           | mean (0.036)          | 0.045               |
|                     | variance (0.00012)    | 0.00035             |
| No-track            | mean (0.042)          | 0.035               |
|                     | variance (0.00020)    | 0.00021             |

Figure 6. (a) Sample semivariogram and (b) sample correlogram for log $\alpha$ with outlier(s) removed in wheel track interrows. Dotted lines indicate 95% confidence limits of autocorrelation. (One interrow count is 76 cm.)

![Figure 6](image)

![Figure 7](image)
Figure 8. Sample semivariogram for log $\alpha$ in Nicollet loam and Clarion loam soils under (a) corn row and (b) no-track interrow. (One row or interrow count is 76 cm.)

central Iowa to compare the effect of corn row, no-track interrow, and wheel track interrow on the spatial variability of soil hydraulic properties ($K$, $K_s$, and $\alpha$). Hydraulic conductivity, measured at 0-, 30-, 60-, and 150-mm water tensions was compared under the three field conditions. Classical ANOVA showed significant treatment difference ($P = 0.01$) between field conditions. For all four tensions, $K$ was largest in the corn rows and smallest in the wheel track interrows, with no-track interrows intermediate. Sample semivariograms estimated for $K$ and $K_s$ under the three field conditions showed some pseudoproportional effect in nugget variance. A clear negative trend (highest variability at lowest tension and lowest variability at highest tension) was observed for no-track interrows. In corn rows this trend was significant but were significantly different from a semivariogram structure across the soil boundary in no-track interrows. Under wheel track interrows, no structure was found, indicating spatial structure was removed by wheel compaction and pore structure destruction.

Acknowledgments. This project was made possible by grant support from the Iowa Department of Natural Resources and The Management Systems Evaluation Area Project. Journal paper J-15399 of Iowa Agriculture and Home Economics Experiment Station, Iowa State University; project number 2715.

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(Received August 16, 1993; revised April 11, 1994; accepted April 15, 1994.)