

Use of Statistical and Spatial Variability Analyses for Investigating Variations of Soil Nitrate in Irrigated Corn Fields

Ramzy M. R. Hedia

Soil & Water Science Department, Faculty of Agriculture, Alexandria University, El-Shatby, Alexandria, Egypt.

THE OBJECTIVES of this study were to examine field variability soil $\text{NO}_3\text{-N}$ content with respect to their management and inherent dependence and to improve sampling strategies of soil $\text{NO}_3\text{-N}$ for future studies on N balance and validation of solute transport models. Two 50x50m plots (control and N-fertilized) in an irrigated corn (*Zea mays* L., hybrid cultivar 10) field were randomly surface (0-15cm) and subsurface (15-30 cm) sampled according to a 2x2m grid system after the first dose of the N fertilizer was applied. Descriptive statistical and spatial variability analysis of soil $\text{NO}_3\text{-N}$, soil moisture and soil organic nitrogen contents were used to determine the different components responsible for field soil $\text{NO}_3\text{-N}$ variability. Significant field variations in soil $\text{NO}_3\text{-N}$ were found within the studied field. Nonuniformity of N-fertilizer application accounted for 84-89% of the total soil $\text{NO}_3\text{-N}$ variability. Soil moisture and soil organic nitrogen contents had significant contribution to soil $\text{NO}_3\text{-N}$ variability in the control plot (57-72%). Geostatistical analysis of the spatial variability indicated that soil $\text{NO}_3\text{-N}$, soil moisture and soil organic N contents of the surface and subsurface layers confirmed to the Gaussian or spherical semi-variogram models. The calculated variograms were used to establish kriging maps of the studied variables. Sampling strategies were determined and were variable between the control and the N-fertilized plots as well as between the surface and subsurface layers.

Most plant-soil systems are spatially heterogeneous and therefore considerable efforts has been invested in quantifying the field scale variability (1-100 m) of biologically important soil properties. Even in fields, presumably homogenized by cultivation, variability can be substantial at small spatial scales (Webster and Butler, 1976; Marriott *et al.*, 1997).

One approach to determine the fate of nitrogen in agricultural systems is by developing N balances (Lund, 1982; Lund *et al.*, 1978; Fried *et al.*, 1976). The most difficult sinks to measure for N balances are leaching and gaseous losses. An important line of research dealt with the intrinsic spatial variability of soil hydraulic properties and transport behavior of salts and inorganic ions (Freebairn *et al.*, 1989; Pikul *et al.*, 1990; Unger, 1992; Poletika and Jury, 1994). Several field-scale studies have shown that movement of water and solutes through unsaturated soil often is highly variable in space even when surface inputs are spatially nearly uniform (Biggar and Nielsen, 1976; Van De Pol *et al.*, 1977; Jury *et al.*, 1982; Tabor *et al.*, 1985). Tabor *et al.* (1984) reported that many factors affect spatial and temporal variability of nitrate content in cropped fields, such as soil properties, cultural practices and physiological maturity. Thus, the interpretation of irregular flow patterns resulting from surface management practices will often be complicated by the large amount of intrinsic variation in transport properties present in field soils (Poletika and Jury, 1994). Validation of the recently used solute transport models to field conditions is also necessary to improve their predictability (Huwe and van der Pleog, 1988; Hutson and Wagenet, 1992; Huwe, 1992; Jabro *et al.*, 1993; Hedia, 2000). Researchers are then faced by spatial and temporal variations in the field solute concentration.

Classical statistical tests assume that observations are independent. However, many soil properties vary continuously over space, with contiguous samples being the most similar. These are, therefore, not independent from each other at some scales. However, geostatistical analyses provide the means to interpret the spatial dependence among soil samples (Burgess and Webster, 1980a; Webster, 1985; Webster 2001). Hence, statistical analysis along with geostatistics can be used as a useful tool to quantify management and inherent spatial dependence of soil moisture and nitrate contents (Burgess and Webster, 1980a, b; Vieira *et al.*, 1981; Russo, 1984; Ramadan and Abdel-Kader, 1995; Ramadan, 2001).

The objectives of this study were (i) to examine field variability of soil moisture, $\text{NO}_3\text{-N}$ and soil organic N contents with respect to their management and inherent dependence and (ii) to improve sampling strategies of soil $\text{NO}_3\text{-N}$ for future studies on N balance and validation of solute transport models.

Material and Methods

1- Study area

This study was conducted on a 2.5 faddan furrow irrigated corn (*Zea mays* L., hybrid cultivar 10) field located in the Experimental Station of the Faculty of Agriculture, Alexandria University at Abis. Corn was sown on 19 May 2001 and harvested on 10 August 2001. The field was chosen because of its apparent uniformity and size. The soil is a sandy clay loam, (Lacustrine) Aquic Torrifluvents. Some selected chemical and physical characteristics of the soil were determined according to the standard methods outlined by Page *et al.* (1982) and are listed in Table 1.

TABLE 1. Some selected chemical and physical characteristics of the studied soil.

Characteristics	Depth (cm)	
	0-15	15-30
pH (1:2.5)	7.68	7.72
E.C. (dSm ⁻¹)	1.29	1.35
CaCO ₃ (g kg ⁻¹)	92.0	111.2
OM (g kg ⁻¹)	25.8	19.5
CEC (cmolc kg ⁻¹) [#]	35.73	31.32
Clay% (<0.002 mm)	32.45	28.37
Textural Class ⁺	SCL	SCL
Ks (m d ⁻¹) ^x	0.097	0.085
F.C. (%v) [*]	41.4	35.9
Bulk density (Mg m ⁻³)	1.29	1.36
Available (mg kg ⁻¹)		
N	1.57	1.07
P	22.6	12.6
K	7.7	3.2

[#] Cation Exchange Capacity.

⁺ SCL= Sandy Clay Loam.

^x Saturated hydraulic conductivity.

^{*} Field capacity on volume basis.

2- Field layout

Two 50x50 m plots were set up in the center of the field with a 5.0 m width buffer zone from all directions. The rows were spaced at 0.70 m apart. One plot received the recommended fertilization rate for corn (N-fertilized plot). Nitrogen was applied as ammonium nitrate fertilizer (33.5% N) at 150 kg/fad., phosphorus was applied as superphosphate fertilizer at 100 kg P₂O₅/fad., and potassium was applied as potash salt (52% K₂O) at 50 kg/fad. The second plot received the same fertilization rates of phosphorus and potassium but did not receive any N fertilizers (control plot). Corn was sown after winter wheat which received 70, 80 and 0 kg/fad. of N, P and K, respectively. The field was conventionally surface-irrigated and common practices were similarly applied to both plots.

3- Soil sampling and laboratory analysis

Using a random-number generator, 50 sites were chosen in each plot from the intersections of a 25x25 regular grid. After the first dose of the N fertilizer, the surface (0-15cm) and subsurface (15-30 cm) layers were randomly sampled according to a 2x2m grid system from both plots. Composite samples were consisted of four soil cores. Samples were collected with a 7.5 cm diameter cores. Samples were packed in polyethylene bags, immediately transported to the laboratory and stored deep-frozen until subsequent analysis carried out.

Collected samples were analyzed for their gravimetric water content (θ) using the loss in weight method. Soil mineral N content (N_{\min} ; $\text{NO}_3\text{-N}$ and $\text{NH}_4\text{-N}$) was extracted using 2.0M KCl and analyzed for $\text{NH}_4\text{-N}$ and $\text{NO}_3\text{-N}$ concentrations. Total soil-N content (N_t) was also analyzed using Vapodust 50 nitrogen distillation unit (Page *et al.*, 1982). Soil organic N content (N_{org}) was calculated as $N_{\text{org}} = N_t - N_{\min}$

4- Methods of data analysis

To study the contribution of the different variation sources to the field distribution of $\text{NO}_3\text{-N}$ content, descriptive statistical analysis of the obtained data was carried out using the SPSS Software package, Version 9.0 (SPSS, 1999). Calculations of the minimum, maximum, range, mean, variance (σ^2), standard deviation (SD) and coefficient of variation (CV%) were evaluated.

Geostatistical analysis of the obtained data was also performed. Geostatistics is based on the theory of regionalized variables. Extensive background material in the soils literature includes articles by Burgess and Webster (1980a,b), Warrick and Nielsen (1980) and Vieira *et al.* (1981). Unlike most classical statistics, the assumption of independence is not made. A variogram function $\gamma(h)$ (which is basic to geostatistics) is defined from

$$2 \gamma(h) = \text{Var} [Z(x) - Z(x + h)] \dots (1)$$

where $Z(x)$ and $Z(x+h)$ are random variables corresponding to sites separated by a vector h .

Mathematical models that are fitted to variograms are useful for subsequent applications, *e.g.* kriging. Valid possibilities include linear, spherical, exponential, gaussian and power models (Warrick *et al.*, 1986; Oliver and Webster, 1991). To date, there is no foolproof, pure objective method for fitting models to sample variograms. As a result, models are fitted subjectively but weighted more heavily on distances for which large number of sample pairs are available and for which pairs are relatively close together.

An idealized, linear variogram model is given in Fig. 1. The semivariogram starts at C_0 for $h = 0$. The limiting "nugget" value C_0 is due to inherent variability of the characteristic type of sampling and/or laboratory analysis error. From C_0 the value increases linearly with distance between samples (h) to a maximum "sill" value ($C_0 + \Delta C$). The semivariance remains constant with intersample distances greater than or equal to the "range" (a). Thus, samples close together have small semivariances and are more alike than samples further apart which have larger semivariance. Samples are dependent for distance up to range "a" where the semivariance then remains constant with increasing distances between samples and samples achieve independence.

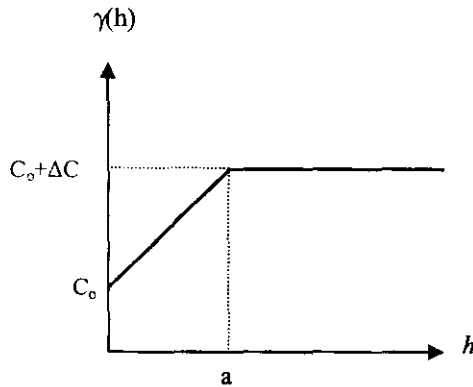


Fig. 1. A linear variogram model (Oliver and Webster, 1991).

Kriging is an optimal linear interpolation method used to predict unknown site values by appropriately weighting the known values on the predicted site through the use of the variogram (Burgess and Webster, 1980a,b). This estimation is unbiased with minimum variance. This method of analysis can be used to prepare maps of predicted site values with a variance for each site. The variance maps can be used to determine where more sampling sites could be located to improve overall estimates.

Results and Discussion

1- Analysis of variance

One of the initial causes of soil NO₃-N variability in a field soil is due to the nonuniform addition of fertilizers (Cameron *et al.*, 1979). The total variance of NO₃-N content due to fertilizer application (σ_T^2) can be calculated by assuming that variation on the N-fertilized plot was due to inherent variation (σ_h^2) already present before fertilization (as measured by the control plot) and variation caused by the nonuniform application of fertilizer (σ_F^2) namely;

$$\sigma_T^2 = \sigma_h^2 + \sigma_F^2 \dots\dots (2)$$

The percent of the variability due to fertilizer application is thus calculated as

$$\% \sigma_F^2 = \frac{(\sigma_T^2 - \sigma_h^2)}{\sigma_T^2} \times 100 \dots (3)$$

Descriptive statistical analysis of soil NO₃-N content (kg/fad./15cm) variability is presented in Table 2. In general, the minimum, maximum, range, mean and SD values of NO₃-N content in the N-fertilized plot were higher than those in the control plot. Similarly, these parameters had higher values in the

surface layers than in the subsurface layers for both plots. On the other hand, the CV% values took the reverse order for plots and layers.

Calculated values of the total variance (σ_T^2) of the field variability of $\text{NO}_3\text{-N}$ content in the N-fertilized plot were 161.7 and 58.0 (kg/fad./15cm)² in the surface and subsurface layers, respectively. The inherent variability of the soil $\text{NO}_3\text{-N}$ content is represented by the variance of the control plot (σ_h^2); 18.0 and 8.9 (kg/fad./15cm)² in the surface and subsurface layers, respectively. The percent variance of $\text{NO}_3\text{-N}$ content due to nonuniformity of N-fertilizer application ($\% \sigma_F^2$) was calculated using Eq. (3). The results in Table 2 show higher $\% \sigma_F^2$ value (88.9%) in the surface layer of the N-fertilized plot than in the subsurface layer (84.6%). As the fertilizer dissolves and ions disperse, one might expect a small decrease in the variation between the surface and subsurface layers (Cameron *et al.*, 1979; Poletika and Jury, 1994). This means that percent variability of $\text{NO}_3\text{-N}$ content due to nonuniformity of N-fertilizer application accounted for > 84% of the total variability. These findings are in a good agreement with those obtained by Cameron *et al.* (1979), Lund (1982) and Poletika and Jury (1994). Since the present work was conducted on a corn cultivated field, the obtained values of variation in $\text{NO}_3\text{-N}$ content due to nonuniformity of N-fertilizer application were lower than those obtained by Cameron *et al.* (1979) for bare soils (>98%). This might indicate that the presence of plants added other sources of variation to the total variance.

TABLE 2. Descriptive statistics of soil $\text{NO}_3\text{-N}$ content (kg/fad./15cm) in the surface and subsurface layers for the control and N-fertilized plots.

Statistics	0-15 cm		15-30 cm	
	Control	N-Fertil.	Control	N-Fertil.
Minimum	2.2	23.0	1.7	4.7
Maximum	19.8	62.1	13.1	36.9
Range	17.6	39.1	11.5	32.2
Mean	9.2	38.4	6.6	20.3
SD	4.2	12.7	3.0	7.6
CV%	46.3	33.1	45.0	37.4
σ_T^2	---	161.7	---	58.0
σ_h^2	18.0	---	8.9	---
$\% \sigma_F^2$	---	88.9	---	84.6

TABLE 3. Descriptive statistics of volumetric soil moisture content (%) and soil organic nitrogen content (kg/fad./15cm) of the surface and subsurface layers for the control plot.

Statistics	(θ)		N_{org}	
	0-15	15-30	0-15	15-30
Minmum	0.21	0.32	691.0	401.1
Maximum	0.57	0.59	2308.0	1703.3
Range	0.36	0.27	1617.0	1302.2
Mean	0.36	0.46	1428.5	997.7
SD	0.11	0.06	417.7	319.7
CV%	33.1	25.2	29.2	32.0
σ^2	1.2E-02	3.17E-3	1.7E+05	1.0E+05

A further detailed statistical analysis of volumetric soil moisture (θ) and soil organic nitrogen (N_{org}) contents of the control plot (no N-Fertilizer applied) are presented in Table 3. In this plot, the residual N_{min} and mineralization of the soil organic N-pool are the main sources of NO_3 -N. Hansen and Jensen (1995) reported that variability in NO_3 -N concentration in such plots are mainly due to (i) variation of sources (residual NO_3 -N and mineralization) and (ii) variation of sinks (leaching, uptake, denitrification, etc.). The results showed that the mean volumetric soil moisture content (θ) in the surface layer (0.36%) was lower than that in the subsurface layer (0.46%). However, the variation in (θ) distribution was higher ($CV\%= 33.1$) in the surface layer than in the subsurface layer ($CV\%= 25.2$). On contrast, the mean N_{org} in the surface layer (1428.5 kg/fad./15cm) was higher than that in the subsurface layer (997.7 kg/fad./15cm). N_{org} was less variable in the surface layer ($CV\%=29.2$) compared with the subsurface layer ($CV\%= 32.0$). As soil moisture content and soil organic nitrogen content are independent, it can be said that the variation in soil moisture and organic nitrogen contents were responsible for > 62% and >57% of the NO_3 -N variability in the surface and subsurface layers, respectively. Other sources of variation such as plant distribution, preferential flow of water and NO_3 -N through cracks, can also have substantial contribution which needs further future investigations.

2- Geostatistical analysis

The 3D-distribution of soil nitrate, soil moisture and soil organic nitrogen contents using the Surfer Software Inc., Version 6.1 (Golden Software, 1995) are presented in Fig. 2-9. Spatial variability analysis was carried out using the GSPLUS software (Gamma Design, 1991). Parameters of the obtained semi-variogram and the best fitted models to these variables are listed in Table 4.

The Gaussian model were the best fitted model to describe the spatial variability of soil nitrate content in the surface (0-15 cm) and subsurface (15-30 cm) layers of the N-fertilized and control plots ($R^2 = 0.864$ and 0.921 ; 0.948 and

TABLE 4. Semi-variogram parameters and fitted models of soil nitrate, moisture and organic nitrogen contents.

Variables	Model	C_0	$C_0 + C$	a	R^2
<u>Nitrate</u>					
Fertilized Plot: 0-15 cm 15-30 cm	Gaussian	127.81	181.80	35.95	0.864
	Gaussian	40.03	65.96	35.14	0.921
Control Plot: 0-15 cm 15-30 cm	Gaussian	13.63	19.17	26.23	0.948
	Gaussian	7.18	9.85	35.82	0.997
<u>Moisture Content</u>					
Control Plot: 0-15 cm 15-30 cm	Gaussian	0.093	0.012	17.03	0.955
	Spherical	0.001	0.003	12.26	0.763
<u>Organic Nitrogen</u>					
Control Plot: 0-15 cm 15-30 cm	Gaussian	1.40E+05	2.11E+05	33.83	0.953
	Spherical	1.02E+05	1.50E+05	74.98	0.643

0.997, respectively). The same trend was also observed with the soil moisture and soil organic nitrogen contents of the surface layer in the control plot ($R^2 = 0.955$ and 0.953 , respectively). However, data of soil moisture and soil organic nitrogen contents of the subsurface layer in the control plot confirmed better to the Spherical model ($R^2 = 0.763$ and 0.643 , respectively) than the Gaussian model. Kriging maps of the studied variables were also established (Fig. 2-9).

In general, the "nugget" variance (C_0) of the three variables studied in the surface layer (0-15 cm) of the N-fertilized and control plots were always larger than those in the subsurface layer (15-30). This can indicate higher inherited spatial variability of these variables in the surface layer than in the subsurface layer for both plots. Similarly, the "sill" variance ($C_0 + \Delta C$), which illustrates the structural variability of the studied variables, was also larger in the surface layers than for the subsurface layers. The spatial dependence of a variable over a specific lag distance is described by the "range" (a). Sampling the soil at a distance less than or equal to " a " comprises a spatial-dependent variability of the measured variables. The best fitted models to the studied variables showed considerable differences in the range values of the soil nitrate content of the surface layer of the N-fertilized and control plots (35.95 m and 26.23 m, respectively). The difference in range values between the surface and subsurface layers of the N-fertilized plot was negligible (0.81 m), whereas this difference was considerable in the control plot (9.58 m). The readily soluble NH_4NO_3 fertilizers made it easy for the nitrate to move to the subsurface layer in the N-fertilized plot. On the other hand, nitrate content in the control plot is highly dependent on the mineralization of the soil organic pool which had its own spatial variability. It can be noticed that the range values of the soil nitrate and organic nitrogen contents of subsurface layer in the control plot (35.82 and 74.98

m, respectively) is higher than those of the surface layer (26.23 and 33.83 m, respectively). It can also be noticed that range values of the soil moisture content of the surface and subsurface layer in the control plot (17.03 and 12.26 m, respectively) is the lowest compared with those for soil nitrate and organic nitrogen contents.

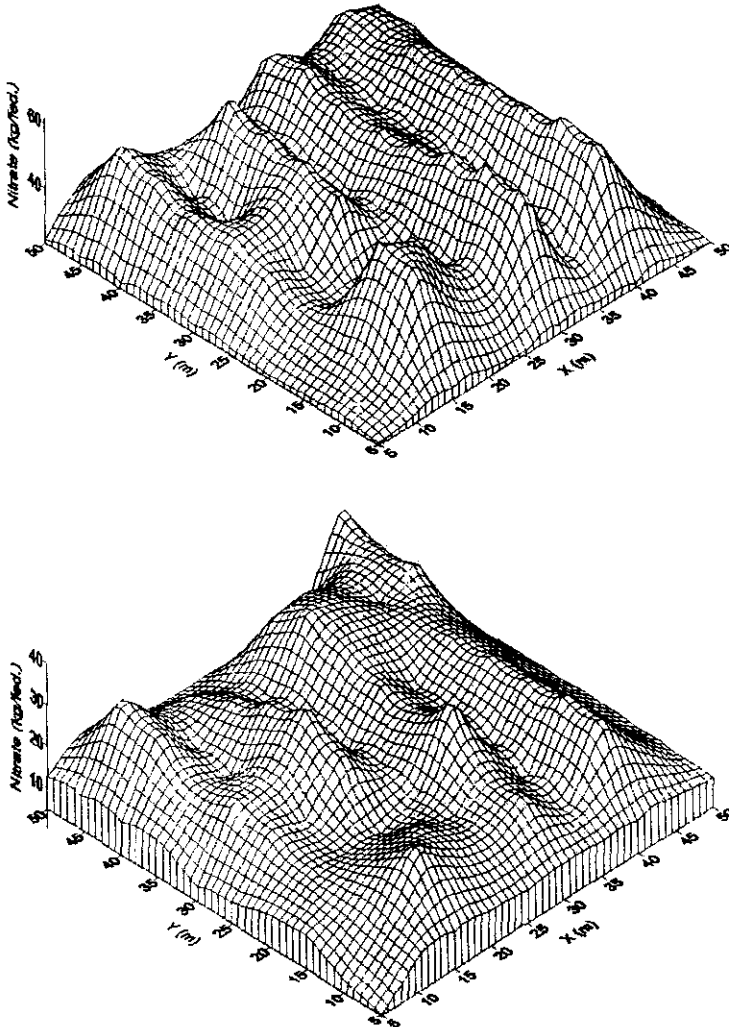


Fig. 2. 3-D distribution of soil nitrate content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the N-fertilized plot.

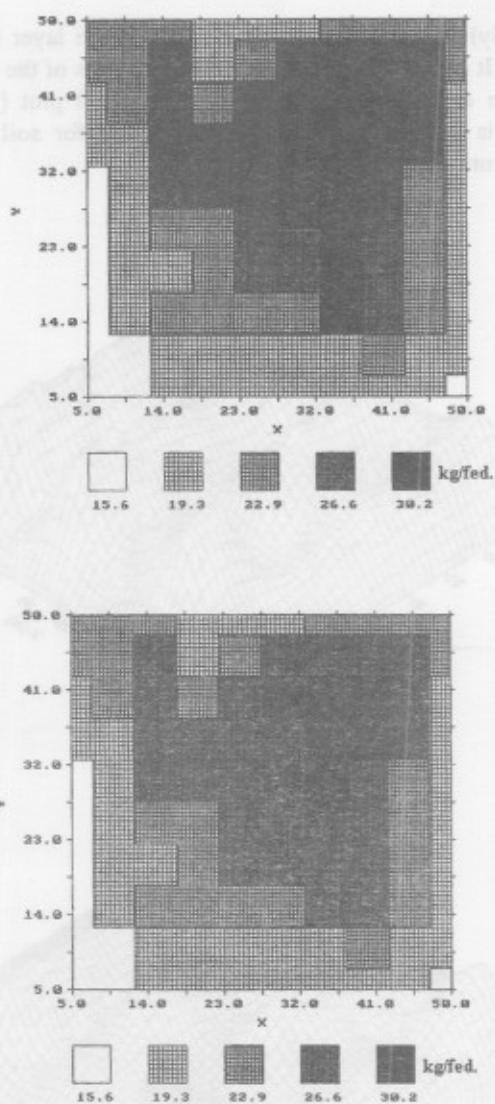


Fig. 3. Kriging maps of soil nitrate content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the N-fertilized plot.

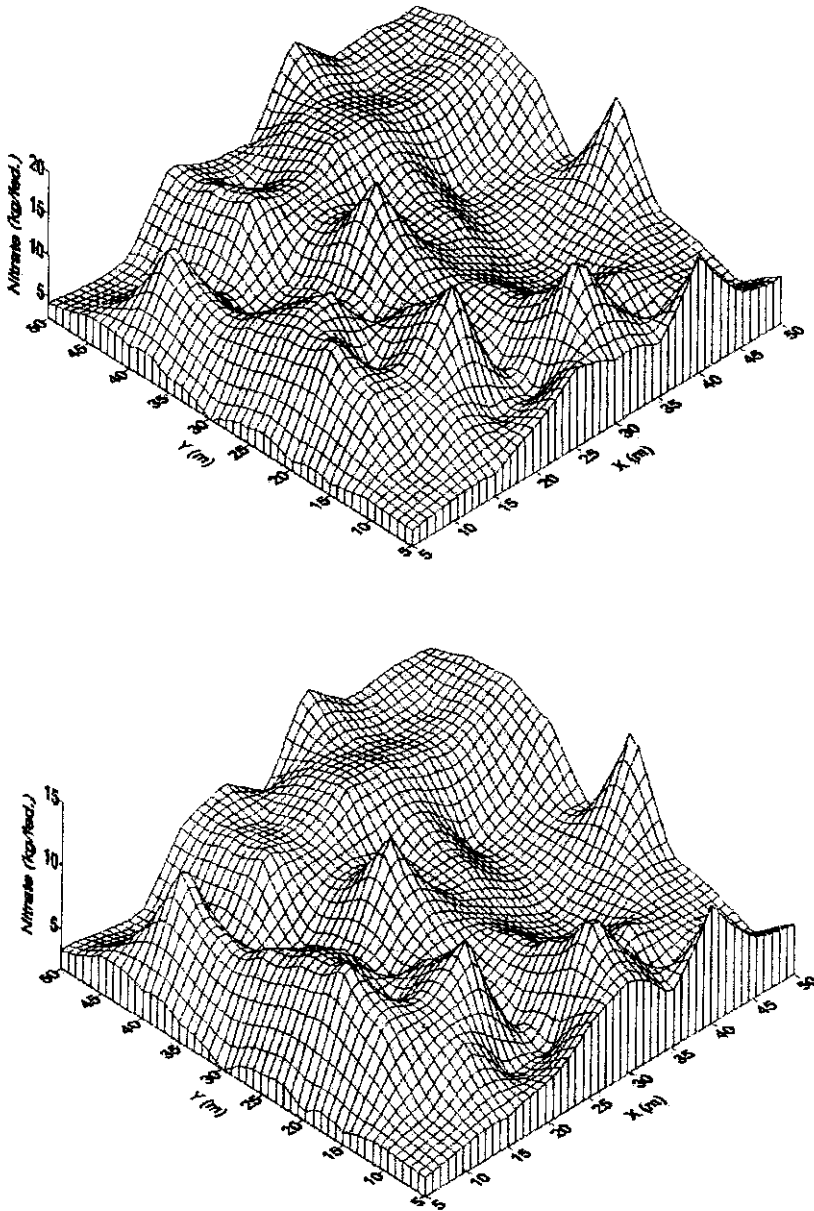


Fig. 4. 3-D distribution of soil nitrate content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

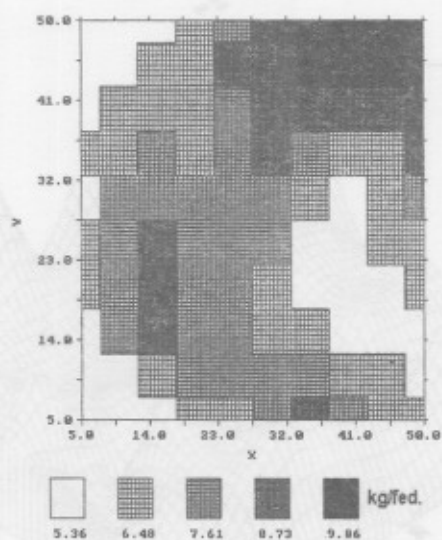
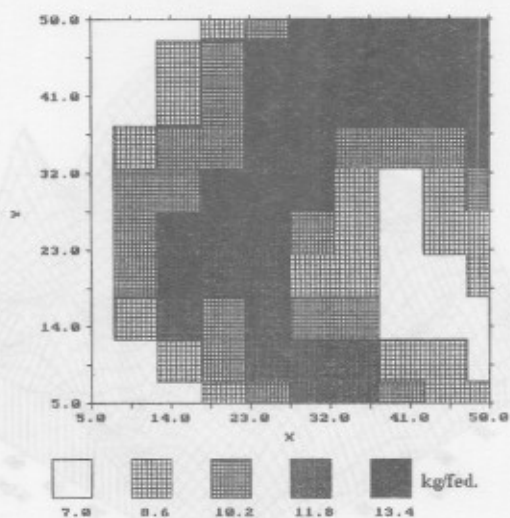


Fig. 5. Kriging maps of soil nitrate content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

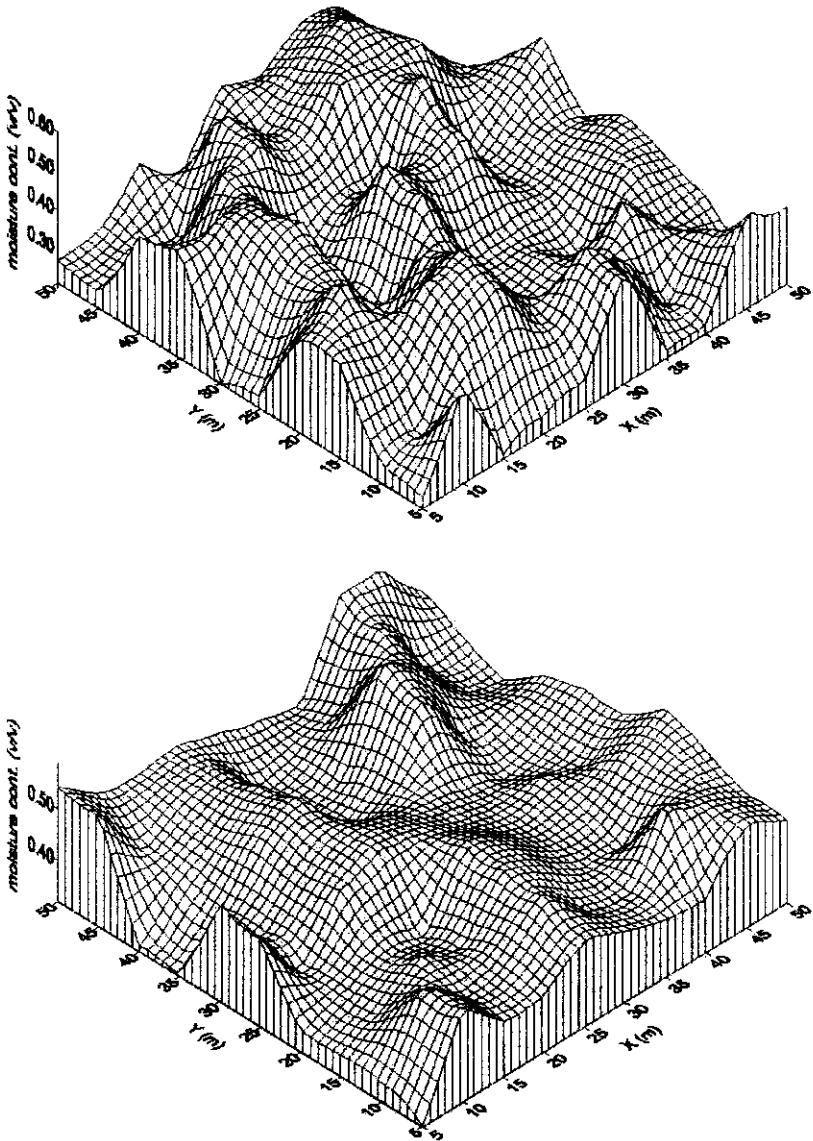


Fig. 6. 3-D distribution of soil moisture content (v/v) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

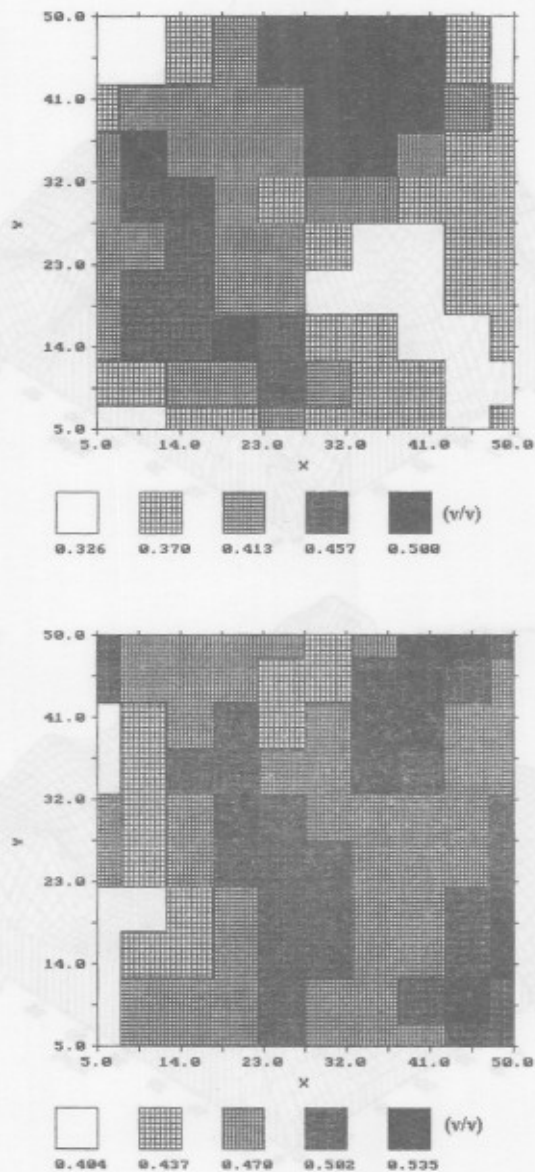


Fig. 7. Kriging maps of soil moisture content (v/v) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

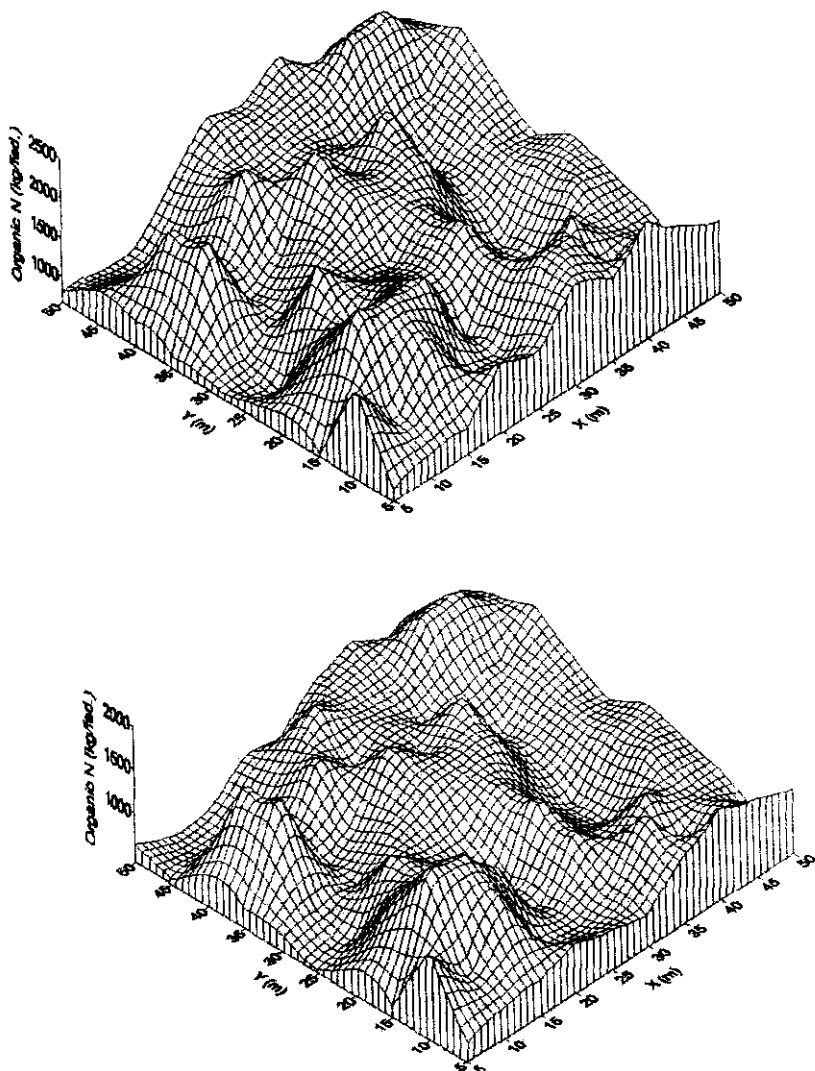


Fig. 8. 3-D distribution of soil organic nitrogen content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

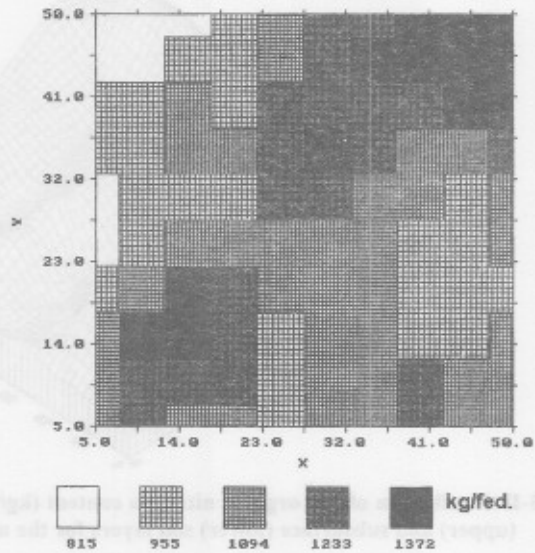
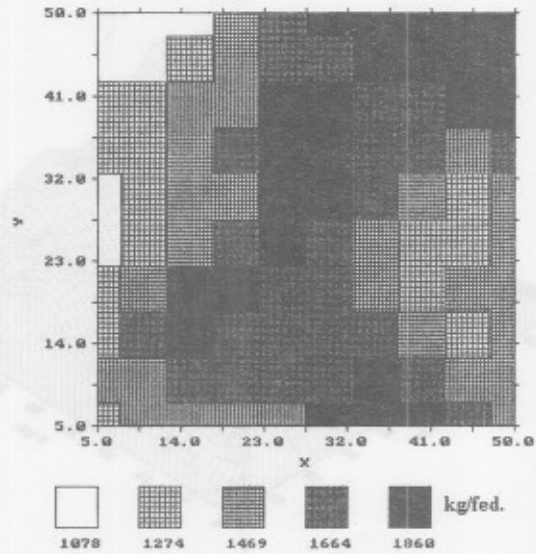


Fig. 9. Kriging maps of soil organic nitrogen content (kg/fed./15 cm) in the surface (upper) and subsurface (lower) soil layers for the unfertilized plot.

Conclusion

One of the initial causes of field soil $\text{NO}_3\text{-N}$ variability is due to the nonuniform addition of fertilizers. Descriptive statistical analysis revealed that the total variance of soil $\text{NO}_3\text{-N}$ content due to fertilizer application can be calculated by assuming that variation on the N-fertilized plot is due to inherent variation already present before fertilization, which can be measured by the control plot and variation caused by the nonuniform application of fertilizer. Analysis of variance revealed that the percent variability of soil $\text{NO}_3\text{-N}$ content under irrigated corn due to nonuniformity of N-fertilizer application accounted for > 84% of the total variability. As soil moisture and soil organic nitrogen contents are independent, the variation in soil moisture and organic nitrogen contents were responsible for >62% and >57% of the soil $\text{NO}_3\text{-N}$ variability in the surface and subsurface layers, respectively. Other sources of variation such as plant distribution, preferential flow of water and $\text{NO}_3\text{-N}$ through cracks, can also have substantial contribution which needs further future investigations. Further geostatistical analysis of the measured variables gave an insight into the spatial dependence of soil $\text{NO}_3\text{-N}$. Such analysis indicated higher inherited and structural spatial variability of soil $\text{NO}_3\text{-N}$, soil moisture and soil organic nitrogen contents in the surface layer than in the subsurface layer for both plots. Sampling lag distances of the studied variables were also evaluated. The integration between descriptive statistics and spatial variability analyses proved to be a good approach to understanding field variability of such variables.

The types of variation frequently encountered in the field situation tend to be complicated. The degree of difference observed in nitrate patterns is often great that the meaning of an average is difficult to comprehend. It is obvious that average soil $\text{NO}_3\text{-N}$ content can mask individual differences in the field. Similarly and when model simulation is in scope, average model parameters and the assumption of uniformity of water and solute distribution mask point to point differences. This fact is often overlooked in examining variable $\text{NO}_3\text{-N}$ distribution in soil profiles and deriving flux boundaries for transport models.

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استخدام طرق التحليل الإحصائي وتحليل الاختلافات المكانية لدراسة الاختلافات في محتوى الأرض من النترات في حقول الذرة المروية

رمزى مرسى رزق هدية

قسم علوم الأراضي والمياه - كلية الزراعة (الشاطبي) - جامعة الإسكندرية -
الإسكندرية - مصر.

تتميز النظم الزراعية الحيوية، وخاصة نظام الأرض مع النباتات بوجود قدر كبير من التباين في خواصها المميزة. فعند دراسة سلوك وتوازن النيتروجين في الحقل باستخدام النماذج الرياضية Models في التوقع بكميات النترات المفقودة بالحقل عادة ما يتم الاعتماد على قيمة وحيدة أو متوسط لمجموعة محدودة من القياسات لصفات الأرض. في حين أنه يجب الأخذ في الاعتبار تأثير الاختلافات في توزيع هذه الصفات في الحقل لوصف النظام بصورة أقرب ما تكون إلى الواقع. وقد كان لهذه الدراسة هدفان. الأول هو دراسة التباين Variation في محتوى الأرض من النترات وعلاقته بعدم انتظام إضافة السماد النيتروجيني والتباين في محتوى الأرض من الرطوبة والنيتروجين العضوي، وذلك عن طريق استخدام طرق التحليل الإحصائي الوصفي Descriptive Statistical Analysis لتحديد مكونات النظام المسنولة عن التباين في محتوى الأرض من النترات، وكذلك استخدام وسائل التحليل الإحصائي الأرضية Geostatistics لدراسة التغيرات المكانية Spatial Variability Analysis للصفات المدروسة. والثاني هو الاستعانة بنتائج هذه التحليلات في وضع استراتيجية لعملية أخذ عينات الأرض Sampling Strategy لتقدير محتوى الأرض من النترات والرطوبة والنيتروجين العضوي في الدراسات المستقبلية.

وقد تم تنفيذ التجربة الحقلية في أرض المزرعة البحثية لكلية الزراعة-جامعة الإسكندرية بمنطقة أبيس. حيث تم زراعة محصول الذرة (Zeamaize L., hybrid cultivar 10) على مساحة ٢,٥ فدان. قسمت هذه المساحة إلى شريحتين بأبعاد ٥٠x٥٠ متر. تم تسميد أحد الشريحتين بالمستويات الموصى بها من عناصر NPK. والأخرى تم تسميدها بعنصر PK فقط (Control). وبعد إضافة الدفعة الأولى من السماد النيتروجيني تم أخذ عينات أرض سطحية (صفر-١٥ سم) وتحت سطحية (١٥-٣٠ سم) بطريقة عشوائية وذلك بنظام شبكة Grid System من رؤس مربعات بأبعاد ٢x٢ متر. وتم تحليل محتوى العينات من الرطوبة والنيتروجين المعدني (نترات وأمونيوم) والنيتروجين الكلي، وتم حساب محتوى الأرض من النيتروجين العضوي.

وقد دلت نتائج التحليل الإحصائي الوصفي على وجود تباين كبير في توزيع النترات في الحقل المدروس. ومن خلال مقارنة التباين لكل من الشريحة المسمدة بالنيتروجين (σ_T^2) والأخرى الغير مسمدة به (σ_H^2) أمكن حساب النسبة المئوية للتباين في محتوى الأرض من النترات نتيجة لعدم انتظام إضافة السماد النيتروجيني (σ_T^2/σ_H^2 %). ووجد أن عدم انتظام توزيع السماد النيتروجيني هو المسئول عن ٨٤-٨٩% من الاختلافات في محتوى الأرض من النترات، وأن التباين في محتوى الأرض من الرطوبة والنيتروجين العضوي ساهمت بنسبة ٥٧-٧٢% من الاختلافات في محتوى الأرض من النترات في الشريحة الغير مسمدة بالنيتروجين. ومن خلال دراسة الاختلافات المكانية لمحتوى الأرض من النترات والرطوبة والنيتروجين العضوي أمكن وصف تغيرها مع المسافة باستخدام النموذج الرياضي من النوع Gaussian في الطبقات السطحية. وكان النموذج الرياضي من النوع Spherical هو الأنسب في وصف تغيرات محتوى الأرض من الرطوبة والنيتروجين العضوي في الطبقات التحت السطحية. وبذلك أمكن تحديد المسافات التي يجب عندها أخذ عينات الأرض من الطبقات السطحية والتحت السطحية في الحقل لتقدير محتوى الأرض من النترات والرطوبة والنيتروجين العضوي. وقد ساعد ذلك في رسم خرائط تقسيمية Kriging Maps للصفات المدروسة.