DEVELOPMENT OF A METHODOLOGY TO DETERMINE THE BEST GRID SAMPLING IN PRECISION COFFEE GROWING

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(Received: January 31, 2018; accepted: April 09, 2018)

ABSTRACT: Precision agriculture is based on a set of techniques that explore the spatial variability of properties related to a determined area. The aim of this study was to develop and test a methodology to evaluate the quality of grid sampling. The experiment was performed in three areas of 112, 50 and 26 ha, in coffee plantations (*Coffea arabica*) with cultivar Catuai 144, in the Três Pontas Farm, located in Presidente Olegário, MG, Brazil, in 2014 and 2015. A total of 224, 100, and 52 georeferenced points (2.0 points/ha) were plotted in the areas regarding the soil chemical properties, respectively: phosphorus, potassium, calcium and magnesium. For the application methodology the standardized accuracy index (SAI), the standardized optimal grid indicator (SOGI) were developed and tested. From grid 1 (2 points/ha), another three sampling grids (1.0, 0.7 and 0.5 point/ha) were adopted. The indexes were important to analyze the grid quality, whereas the SOGI allowed selecting the grid that best represented the properties.

Index terms: Precision agriculture, geostatistics, grids, soil fertility.

DESENVOLVIMENTO DE UMA METODOLOGIA PARA DETERMINAÇÃO DA MELHOR MALHA AMOSTRAL EM CAFEICULTURA DE PRECISÃO

RESUMO: A agricultura de precisão baseia num conjunto de técnicas que explora a variabilidade espacial dos atributos de uma área. O objetivo deste trabalho foi desenvolver e testar uma metodologia para avaliar a qualidade de malhas amostrais. O experimento foi desenvolvido nos anos de 2014 e 2015 na fazenda Três Pontas, Presidente Olegário/MG, em três áreas de 112, 50 e 26 ha, todas de lavoura de cafeeiro (*Coffea arabica*) cultivar Catuai 144. Demarcaram-se nas áreas 224, 100 e 52 pontos georreferenciados respectivamente (2,0 pontos/ha), os atributos químicos do solo testados: fósforo, potássio, cálcio e magnésio. Para aplicação da metodologia foi desenvolvido e testado o índice de exatidão padronizado (IEP), índice de precisão paradronizado (IPP) e o indicador de malha ótima padronizado (IMOP). A partir da malha 1 (2 pontos/ha) foram adotadas mais 3 malhas amostrais (1,0; 0,7 e 0,5 ponto/ha). Os índices mostraram importantes para analisar a qualidade das malhas e o IMOP permitiu a escolha da malha que melhor representou os atributos.

Termos para indexação: Agricultura de precisão, geoestatistica, grids, fertilidade do solo.

1 INTRODUCTION

Coffee is one of the main agricultural products of Brazilian agribusiness. Because it is an activity of great importance for the Brazilian trade balance, there is a need to seek the deep knowledge of all the productive processes involved from planting to harvesting (FERRAZ et al., 2012b).

The coffee growers are in a continuing search for technologies that aid in the management and improve the production of their crop, adapting to the market demands.

Precision agriculture comes as a tool that contributes to cost savings. For Rodrigues Junior et al. (2011), precision agriculture might bring countless benefits to the coffee growing, since it has a high income per area and the main parameter is bean quality. According to Ferraz (2012), precision agriculture in coffee growing has been referred as "precision coffee growing".

Ferraz et al. (2012) define it as a set of techniques and technologies capable of assisting the coffee farmer to manage the crop, based on the spatial variability of soil and plant properties, in order to maximize profitability, increase efficiency of fertilization, spraying and harvesting, thus increasing productivity and the product's final quality.

The understanding on spatial variability in the crop requires a greater amount of information, which can be obtained from sampling operations (SOUZA; MARQUES JÚNIOR; PEREIRA, 2004). These samplings, mainly referring to grid sampling, still generate discussions among scientists, technicians and producers, which do not yet have well-established standards for coffee growing.

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According to Ferraz et al. (2017), the use of unsatisfactory grid sizes can generate maps that do not reflect the field and thus generate wrong technical recommendations, which could result in losses to its users.

According to Nanni et al. (2011), the grid sampling used in the most diverse Brazilian cultures are around one point every two to three hectares and, in some cultures, up to one point is used every four hectares. In relation to the coffee crop, few studies refer to the adequate size of grid sampling, thus, there is a great need for research with this purpose.

The aim in this study was to develop and test a methodology to evaluate and compare the quality of different grid samples in different size areas and define the one that best characterizes the spatial variability for each tested property.

2 MATERIAL AND METHODS

The experiment was performed in three different areas (112 ha, 50 ha and 26 ha) at Três Pontas Farm, Presidente Olegário, MG, Brazil, in 2014 and 2015, all based on wet process (*Coffea arabica* L.) with cultivar Catuaí IAC 144, planted in December 2005 (11-year crop), December 2011 (5 years) and December 2004 (12 years), respectively, spacing of 4.0 m between rows and 0.5 m between plants, totaling 5,000 plants.ha⁻¹.

They were demarcated in the study area, using a GPS signal receiver equipment, 224, 100 and 52 georeferenced sample points (2.0 points/ ha), respectively. From grid 1, another three grid samples were adopted (1.0, 0.7 and 0.5 point/ha).

Using a GPS signal receiver, the demarcated points were found and then the samples collection was made. The grid 1 has the greater number of georeferenced sample points, with 2.0 points/ha, the grid 2 has 1.0 point/ha, the grid 3 refers to 0.7 point/ha and the grid 4 is composed by 0.5 point/ha.

Soil fertility data were collected at every georeferenced point in the area. Each sampling point corresponded to four plants, being that the collection was performed in July 2014 and July 2015 using a quadricycle, by removing subsamples in the projection of coffee canopy at depth from 0 to 20 cm. The samples were sent to the Brazilian Laboratory of Agricultural Analysis (LABRAS), in Monte Carmelo, MG. The chemical attributes of the soil analyzed were Phosphorus, Potassium, Calcium and Magnesium. In order to reduce costs with the collection of soil samples in 2015, in the area of 112 ha, the sampling was done from grid 2, referring to one point per hectare.

Semivariograms are used in order to analyze the spatial dependence of properties under study. Semivariance is classically estimated by equation 1:

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

where, N (h) is the number of experimental pairs of observations $Z(x_i)$ and $Z(x_i + h)$ separated by a distance h. The semivariogram is represented by the graph $\hat{\gamma}(h)$ versus h. From the fit of a mathematical model to the calculated values of $\hat{\gamma}(h)$, the coefficients of the theoretical model were estimated for the semivariogram called nugget effect (C₀); sill (C₀+C₁); and range (a), as described by Bachmaier and Backers (2011).

The weighted least squares method and the spherical, exponential and Gaussian models were used according to the best fit for each property and for tested grids. With the semivariograms ready, ordinary kriging and kriging validation procedures were performed to assess the interpolation quality.

The software used for geostatistical analysis and map generation was ArcGIS 10. The geostatistical index proposed by Seidel and Oliveira (2014) was used to measure the spatial dependence. Where, the spatial dependence is weak whether ratio is lower than 12.5%; moderate, between 12.5% and 25%, and strong spatial dependence when the ratio is greater than 25%. This index can be used for the spherical, exponential and Gaussian semivariogram models, which is expressed by equation 2:

$$IDE_{model(\%)} = MI.\left(\frac{C_1}{C_0 + C_1}\right).\left(\frac{a}{q.MD}\right).100$$
⁽²⁾

where MI is the model index; C_0 , the nugget effect; C_1 , the contribution; a, the practical range; and q.MD, the value corresponding to the fraction (q) reached at maximum distance (MD) between sampled points. Whether the ratio (a/q.MD) results in a value greater than 1, the ratio is then set to 1, so that it assumes only values between zero and 1. Moreover, the greater the MI value, the greater the spatial dependence of model.

One of the ways to evaluate the quality of estimation and fitting of semivariograms, besides other research characteristics, such as the sample mesh, is through validation. It makes possible to extract some useful values for the observation of errors presented by each grid, such as the mean of standard error (M_{se}) (Equation 3), which should have the value closer to zero, and the standard deviation of standard error (SD_{se}) (Equation 4), which should be as lower as possible.

$$M_{(SE)} = \frac{1}{n} \sum_{i=1}^{n} \frac{[\hat{z}(x_i) - z(x_i)]}{\sigma_{k_{(Ni)}}^2}$$
(3)

$$SD_{(SE)} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} \frac{[\hat{z}(x_i) - z(x_i)]^2}{\sigma_{k_{(XI)}^2}^2}}$$
(4)

Where n is the number of data; $\hat{Z}(x_i)$ is a value predicted or estimated by ordinary kriging at point xi, without considering the observation Z(xi); Z(xi) is the value observed at point xi; and $\sigma_{k(xi)}$ is the standard deviation. Based on these values, a methodology

Based on these values, a methodology was developed and tested covering as main characteristic the use of standard errors, which can be applied in several circumstances, such as in the comparison between grids from different cultures, representing a great differential in relation to other methodologies developed previously.

Based on the MSE and SDSE values obtained it was developed and tested the Standardized Accuracy Index (SAI) and the Standardized Precision Index (SPI). These indexes make possible to identify the best grid for areas up to 100 ha. It was used attributes of the soil to test the indexes, being evaluated Phosphorus (P), Potassium (K), Calcium (Ca) and Magnesium (Mg); which were sampled in the years 2014 and 2015.

The value of the MSE, obtained by validation, reflects the accuracy of grid sampling. The SAI concept (Equation 5) was proposed and developed to determine an accuracy component that would allow comparing among grids.

$$SAI = \frac{M_{(SE)}}{mM_{(SE)}}$$
(5)

where $M_{(SE)}$ is the value of the mean of standard error, in module, of the grid to be compared and $mM_{(SE)}$ is the greatest value of the mean of standard error, in module, among all analyzed grids.

The value of the SD_{SE} , obtained by validation, reflects the grid accuracy. Moreover, the SPI was developed and proposed to compare the accuracy component of the grid among the different studied grids (Equation 6).

$$SPI = 1 - \frac{SD_{(SE)}}{mSD_{(SE)}}$$
(6)

where SD_{SE} is the value of the standard deviation of standard error of the studied grid and mSD_{SE} is the greatest value of the standard deviation of standard error presented by the group of analyzed grids. The SAI and SPI values range from zero to one, and the closer to one, the more accurate/precise is the grid sample, while the closer to zero, the more inaccurate/imprecise is the grid sample.

The SOGI was developed and tested in order to choose the best grid among those under study, taking into account the weighting between the standardized accuracy and prediction indices (Equation 7).

$$SOGI = (0.5 \text{ x SAI}) + (0.5 \text{ x SPI})$$
 (7)

In the equation it is calculated a proportion between SAI and SPI indexes, since both accuracy and precision are important for the grid quality index. As the two indicators (accuracy and precision) were considered equally important, it was given equal weight to both, that is, 50% for each one.

The SOGI ranges from zero to one and the closer to one (or 100%), the better the grid (more accurate and more precise), while the closer to zero (0%), the worse (the more inaccurate and imprecise) is the grid.

3 RESULTS AND DISCUSSION

Cherubin et al. (2015) studied the phosphorus (P) property and observed that the increase in the number of collected samples (n), provided by smaller grid samples, makes it possible to identify sites in the area with extreme P levels that, if not corrected, might result in crop restriction zones.

Carvalho (2016) studied the optimal sampling density for precise coffee growing in an area of 22 ha and found that the grid with 2.0 points/ha were the most suitable for productivity and soil fertility of coffee.

By evaluating grid samples for an area of 22 ha, Ferraz et al.(2017) applied another

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methodology and found the 3.0 points per hectare as the most suitable for the soil fertility variables Phosphorus (P), remaining Phosphorus (P-rem), Potassium (K), and changeability of calcium at pH 7.0 (T). The methodology proposed by Ferraz et al. (2017) encompasses the standard error of the data and was studied for smaller areas, up to 22 ha, while the methodology of this work is based on standardized errors and it was studied in larger areas, up to 100 ha. Moreover, the great difference between the two methodologies is that the one used in this work can be applied in study of different cultures, differing from the one applied by Ferraz which can only be used for coffee cultivation.

Comparative studies were also performed to evaluate grid samples for different variables. Cherubin et al. (2014), Nanni et al. (2011) and Ragagnin, Sena Júnior and Silveira Neto (2010) recommend the use of a sampling with more points per hectare.

Table 1 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the properties Phosphorus (P) and Potassium (K) in the years 2014 and 2015 in an area of 112 ha. For the variable P, in 2014, the geostatistical index proposed by Seidel and Oliveira (2014) showed values considered as weak, since they had a percentage lower than 12.5% in all grid samples. The SOGI relates the SAI and SPI indices, ranging from 0 to 100%, indicating that the best grid for Phosphorus was the grid 3 in the year 2014. For the parameters estimated by the semivariogram fitted by the weighted least squares method and by the exponential model of P in 2015, the spatial dependence index (SDI) showed weak spatial dependence for all grids, and the best grid sample in the year 2015 for P was the grid 4, according to SOGI.

For the parameters estimated by the semivariogram fitted by the weighted least squares method and by the spherical and exponential model for the property Potassium (K) in 2014, it is observed that the geostatistical index is presented as weak for all grid samples, i.e., a spatial dependence lower than 12.5%. In contrast, grids 1 and 3 showed values close to 40% for SOGI. The parameters estimated by the semivariogram fitted by the weighted least squares method and by the spherical model for the Potassium variable, for the year 2015. The IDE was weak for grid samples 2 and 3, and moderate for grid 4. Analyzing all the grid samples, the grid 4 showed the most indicated value for the SOGI, with 44%.

Table 2 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the properties calcium (Ca) and Magnesium (Mg) in the years 2014 and 2015 in an area of 112 ha.

For the property Calcium (Ca), the geostatistical index was weak for most of the grids fitted by the weighted least squares method and by the spherical model for 2014 and 2015, and only the grid 3 of the year 2015 was moderate, according to the index. The SOGI showed values of 45.14% in grid 1 and 43.11% in grid 3 for 2014 and 44.96% in grid 2 in the year 2015, being the best values among the grid samples. The parameters estimated by the weighted least squares method and the spherical model for the variable Magnesium (Mg) in the years 2014 and 2015, presented the geostatistical index as weak and moderate for grid samples in both years, respectively. The SOGI presented values of 43.33% for grid 1 and 40.7% for grid 4 in 2014. For the year 2015 the values presented for grids 1 and 4 were of 42.44% and 40.56% respectivelywhich are the closest to 100% for both years.

Table 3 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the properties Phosphorus (P) and Potassium (K) in the years 2014 and 2015, in an area of 50 ha.

For the variable P, in 2014, the geostatistical index presented values considered as moderate for grid 3 and weak for grids 1, 2 and 4, which had percentage lower than 12.5%. Through the SOGI, it was observed that grid 2 was the best grid for Phosphorus in 2014. For the parameters estimated by the semivariogram fitted by the weighted least squares method and by the exponential model of the P in 2015, the SDI showed weak spatial dependence for grids 1 and 2 and moderate for grids 3 and 4. According to SOGI, the grid 4 was the best grid sample for Phosphorus in the year 2015, with 49.48%.

For the property Potassium (K), in the year 2014, the SDI was weak for all grid samples and, according to the SOGI classification, the best grid sample was grid 1, which showed an index value of 36.51%. For the year 2015, still for the variable K, the SDI was weak for all grids, and the greatest SOGI was 51.67% for the grid 1.

Table 4 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the properties calcium (Ca) and Magnesium (Mg) in the years 2014 and 2015 in an area of 50 ha.

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Grid	NPG	C,	\mathbf{C}_1	$C_0 + C_1$	V	SDI	Seidel	CV(%)	$\mathbf{M}_{(\mathrm{SE})}$	${ m SD}_{ m (SE)}$	SAI	IdS	SOGI
						Phosp	ohorus 20	14					
]*	224	2100.03	103.02	2203.05	166.31	0.50	Weak	33.45	0.0119	0.9697	0.5704	0.0260	29.82
2*	112	1395.83	833.87	2229.70	174.12	4.18	Weak	33.03	0.0139	0.9817	0.4982	0.0140	25.61
3*	62	1941.97	0.00	1941.97	1093.17	0.00	Weak	31.22	0.0073	0.9580	0.7365	0.0377	38.71
4*	57	2008.40	0.00	2008.40	434.93	0.00	Weak	31.54	0.0277	0.9956	0.0000	0.0000	0.00
						Phosp	ohorus 20	15					
2**	112	536.18	615.51	1151.69	228.23	6.62	Weak	25.87	0.0084	1.0118	0.5307	0.0180	27.43
3**	62	527.63	490.51	1018.14	176.39	4.62	Weak	25.75	0.0179	1.0303	0.0000	0.0000	0.00
4**	57	805.02	193.85	998.87	246.24	2.61	Weak	26.07	0.0015	1.0086	0.9162	0.0211	46.86
						Pota	ssium 201	4					
	224	1516.27	343.21	1859.48	167.95	1.99	Weak	33.29	0.0061	0.9796	0.8365	0.0000	41.82
2*	112	1476.09	808.29	2284.38	522.04	11.87	Weak	33.48	0.0122	0.9562	0.6729	0.0239	34.84
3*	62	640.22	1508.17	2148.39	161.53	7.31	Weak	34.63	0.0096	0.9125	0.7426	0.0685	40.56
4**	57	1897.91	0.00	1897.91	258.26	0.00	Weak	37.02	0.0373	0.9555	0.0000	0.0246	1.23
						Pota	ssium 201	2					
2*	112	1035.23	854.72	1889.95	281.92	8.18	Weak	65.58	0.0077	0.9748	0.6435	0.0000	32.18
3*	62	1677.76	242.76	1920.52	289.83	2.35	Weak	68.60	0.0216	0.9429	0.0000	0.0327	1.64
4*	57	0.00	1351.08	1351.08	264.97	17.09	Mod	59.26	0.0025	0.9637	0.8843	0.0114	44.78
NPG - Nt - Moderat accuracy	e; Weak ndex: SP	points of the - Weak; CV I - Standardi	e grid sampl (%) - Coefi ized precisio	ing; C ₀ - Nug ficient of var m index: SO(get effect; ¹ iation; M _{(SE} Al - Standar	C ₁ - Contr) - Mean rd optimal	ribution; C of standar I <u>erid indi</u>	C ₀ +C ₁ - Sil d error; S cator.	l; a - range; D _{se} - Standa	SDI - Spat ard deviatic	ial dependen n of standar	ce index; Str d error; SAI ·	- Strong; Mod · Standardized
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Crid	NPG	Ĵ	ບີ	$C_{0} + C_{1}$	а	SDI	Seidel	CV(%)	$\mathbf{M}_{(SE)}$	$SD_{(SE)}$	SAI	IdS	SOGI
							Calcium 20)14					
1*	224	1.08	0.06	1.14	296.00	1.08	Weak	64.66	0.0031	0.9690	0.8495	0.0532	45.14
2*	112	0.89	0.65	1.54	208.14	5.62	Weak	69.22	0.0056	0.9720	0.7282	0.0503	38.92
3*	79	0.60	1.40	2.00	214.02	9.65	Weak	74.91	0.0039	0.9707	0.8107	0.0516	43.11
4*	57	1.69	0.00	1.69	390.83	0.00	Weak	71.57	0.0206	1.0235	0.0000	0.0000	0.00
							Calcium 20	15					
2*	112	1.51	0.28	1.79	1583.29	15.92	Weak	60.06	0.0036	1.0235	0.8992	0.0000	44.96
3*	79	1.41	0.38	1.79	1571.7	13.32	Mod	59.20	0.0357	1.0197	0.0000	0.0037	0.19
4*	57	0.74	0.68	1.42	246.24	7.60	Weak	63.34	0.0098	1.0005	0.7255	0.0225	37.40
						Z	lagnesium (2014					
1*	224	0.50	0.15	0.65	313.12	4.65	Weak	65.89	0.0031	0.9775	0.8622	0.0044	43.33
2*	112	0.46	0.30	0.76	266.78	6.80	Weak	66.86	0.0225	0.9708	0.0000	0.0112	0.56
3*	62	0.52	0.43	0.95	267.17	7.71	Weak	71.72	0.0196	0.9443	0.1289	0.0382	8.35
4*	57	0.37	0.35	0.72	258.26	8.15	Weak	64.34	0.0042	0.9818	0.8133	0.0000	40.67
						N	lagnesium .	2015					
2*	112	0.35	0.12	0.47	1583.29	12.99	Mod	77.98	0.0036	1.0475	0.8487	0.0000	42.44
с, ж	62	0.34	0.03	0.37	176.40	0.92	Weak	79.19	0.0238	1.0180	0.0000	0.0282	1.41
4*	57	0.26	0.06	0.32	530.25	6.41	Weak	81.34	0.0055	1.0032	0.7689	0.0423	40.56

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Grid	NPG	C ₀	C_1	$C_0 + C_1$	A	SDI S	Seidel	CV(%)	$\mathbf{M}_{(\mathrm{SE})}$	$SD_{(SE)}$	SAI	SPI	SOGI
						Pho	sphorus 2	2014					
1*	100	900.11	553.15	1453.26	120.48	3.50	Weak	49.68	0.0193	1.0116	0.5896	0.0455	31.76
2*	50	881.78	885.71	1616.41	658.95	12.19	Weak	44.28	0.0150	1.0318	0.6809	0.0264	35.36
3**	33	782.41	1166.08	1856.10	770.09	16.39	Mod	42.69	0.0333	1.0598	0.2915	0.0000	14.57
4*	25	1287.53	0.00	1287.53	1058.60	0.00	Weak	41.58	0.0470	1.0313	0.0000	0.0269	1.34
						Pho	sphorus 2	2015					
* *	100	247.47	203.37	450.84	167.56	5.77	Weak	43.09	0.0115	0.9871	0.6338	0.1016	36.77
2*	50	365.64	150.74	516.38	910.77	10.62	Weak	42.99	0.0119	1.0785	0.6210	0.0184	31.97
с, ж	33	381.34	264.56	645.90	993.74	16.31	Mod	45.47	0.0314	1.0987	0.0000	0.0000	0.00
4*	25	295.88	327.65	623.53	705.73	14.81	Mod	46.39	0.0009	1.0787	0.9713	0.0182	49.48
						Pot	tassium 2	014					
1*	100	1185.50	729.43	1914.93	386.94	11.25	Weak	32.09	0.0103	1.0379	0.6971	0.0332	36.51
2*	50	1440.04	500.16	1940.20	1092.93	11.25	Weak	28.99	0.0289	1.0665	0.1500	0.0065	7.83
ж	33	1523.92	191.04	1714.96	287.65	2.57	Weak	31.20	0.0340	1.0735	0.0000	0.0000	0.00
4*	25	1692.79	0.00	1692.79	705.73	0.00	Weak	30.73	0.0214	1.0328	0.3706	0.0379	20.42
						Pot	tassium 2	015					
1*	100	159.29	439.32	598.61	134.91	7.55	Weak	46.04	0.0092	0.9539	0.8999	0.1334	51.67
2*	50	482.20	86.29	568.49	176.40	2.14	Weak	46.93	0.0306	0.9918	0.6670	0660.0	38.30
3*	33	626.58	76.05	702.63	203.50	1.76	Weak	51.33	0.0079	1.0264	0.9140	0.0676	49.08
4*	25	122.17	203.09	325.26	213.44	10.64	Weak	42.02	0.0919	1.1008	0.0000	0.0000	0.00
- Moder	Number (rate; Wea	of points of k - Weak; (the grid sa CV (%) - C	mpling; C ₀ - Coefficient of	Nugget effec f variation; N	tt; C ₁ - Co 1 _(SE) - Mea	ntribution in of stand	; C ₀ +C ₁ - S dard error;	Sill; a - ran SD _{SE} - Sta	ge; SDI - Sp ndard devia	atial depend tion of stand	lence index; St lard error; SAI	r - Strong; Mod - Standardized
accurac	y muex; ;	SF1 - Stanus	arurzeu pre	cision index;		uaru opum	nai griu in	ulcator.					

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by the spherical (*) and Gaussian model (***) for the	
east squares method l	4 and 2015.
by the weighted l	na in the years 201
e semivariogram fitted	(Mg) in the area of 50]
rameters estimated by the	um (Ca) and Magnesium
TABLE 4 - Pai	oroperties Calci

Grid	NPG	C	c	$C_0 + C_1$	A	SDI	Seidel	CV(%)	$\mathbf{M}_{(\mathrm{SE})}$	${ m SD}_{ m (SE)}$	SAI	SPI	SOGI
					C	alcium 2	014						
1*	100	1.53	0.17	1.70	1137.16	4.40	Weak	42.00	0.0068	0.9932	0.5310	0.0259	27.85
2*	50	1.66	0.18	1.83	1092.93	4.18	Weak	38.95	0.0057	1.0196	0.6069	0.0000	30.34
3*	33	1.78	0.34	2.13	993.74	6.40	Weak	39.22	0.0066	0.9820	0.5448	0.0369	29.09
4*	25	1.82	0.17	2.00	1058.60	3.66	Weak	40.92	0.0145	0.9882	0.0000	0.0308	1.54
					C	alcium 2	015						
1*	100	0.96	1.55	2.51	154.07	7.26	Weak	47.80	0.0009	0.9715	0.9833	0.0506	51.70
2*	50	0.83	1.44	2.27	197.23	9.99	Weak	46.02	0.0182	1.0074	0.6630	0.0155	33.93
3*	33	2.01	0.01	2.02	229.56	0.09	Weak	43.39	0.0540	0.9265	0.0000	0.0946	4.73
4*	25	1.80	0.00	1.80	372.15	0.00	Weak	46.07	0.0134	1.0233	0.7519	0.0000	37.59
					Ma	Ignesium	2014						
	100	0.31	0.06	0.37	159.02	2.00	Weak	73.09	0.0158	0.9952	0.5798	0.1017	34.08
2*	50	0.32	0.07	0.39	235.97	3.17	Weak	71.90	0.0047	1.0246	0.8750	0.0752	47.51
3*	33	0.13	0.30	0.44	224.47	12.49	Weak	71.57	0.0376	1.0024	0.0000	0.0952	4.76
4***	25	0.05	0.49	0.54	223.86	21.79	Mod	76.12	0.0131	1.1079	0.6516	0.0000	32.58
					Ma	Ignesium	2015						
1	100	0.25	0.18	0.43	161.41	5.16	Weak	67.64	0.0122	0.9656	0.8082	0.0724	44.03
2*	50	0.25	0.16	0.41	185.00	5.77	Weak	68.77	0.0222	1.0410	0.6509	0.0000	32.55
3*	33	0.33	0.00	0.33	418.25	0.00	Weak	65.16	0.0636	1.0264	0.0000	0.0140	0.70
4*	25	0.38	00.00	0.38	1058.60	0.00	Weak	69.72	0.0103	1.0276	0.8381	0.0129	42.55
NPG - Number - Moderate; We	of points of alk - Weak; C	The grid sar TV (%) - C	npling; C ₀ oefficient	- Nugget eff of variation;	eet; C ₁ - Cc M _(SE) - Me:	an of star	n; C ₀ +C ₁ ndard errc	- Sill; a - r or; SD _{SE} - 1	ange; SDI Standard de	- Spatial d	ependence standard	error; SAI -	- Strong; Moc Standardizec
accuracy index;	SP1 - Standa	urdized prec	cision indec	x; SUGI - Du	andard opu	nal grid 1	ndicator.						

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For the property Calcium (Ca), the geostatistical index was weak for all the grids fitted by the weighted least squares method and by the spherical model for 2014 and 2015. The SOGI showed greater values in grids 2 (2014) and 1 (2015), being 30.34% and 51.70%, respectively.

For the variable Magnesium (Mg), the geostatistical index was weak for all the grids of the years 2014 and 2015, except for the grid 4 of the year 2014, where the index was moderate. The SOGI presented values of 47.51% for grid 4 in 2014, and 44.03% and 42.55% for grids 1 and 4, respectively, in the year 2015, which are the closest to 100%.

Table 5 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the properties Phosphorus (P) and Potassium (K) in the years 2014 and 2015 in an area of 26 ha.

For the property Phosphorus (P), the geostatistical index was moderate for the grid 1 and weak for the other grids in the year 2014. For the year 2015, the SDI was moderate for grids 1 and 2 and weak for grids 3 and 4. The SOGI showed values of 38.44% in grid 2 and 37.33% in grid 3 for 2014 and 32.79% in grid 3 in the year 2015, being the best values among the grid samples. For the variable Potassium (K), the

geostatistical index was moderate for grid 1 and weak for grid 2, 3, and 4, in the year 2014. For the year 2015, the SDI for K was weak for grids 1 and 3, moderate for grid 3, and strong for grid 4. The SOGI presented values of 35.90% for grid 1 in the year 2014 and 55.04% for grid 1 in the year 2015.

Table 6 presents the parameters estimated by the semivariogram fitted by the weighted least squares method for the Calcium (Ca) and Magnesium (Mg) in the years 2014 and 2015 in an area of 26 ha.

For the property Magnesium (Mg), the geostatistical index was moderate for grid 3 and weak for grid 1, 2 and 4 in the year 2014. For the year 2015, the SDI was weak for all grids. The SOGI presented values of 46.71% for grid 1 in the year 2014 and 38.07% for grid 4 in the year 2015.

Based on Table 7, the grids considered as more adequate for the area of 112 ha were grids 1 and 2, not statistically different from each other, by the Scott Knott Test, at 5% probability. Grid 1 is more recommended for areas of 50 ha and 26 ha, followed by grid 2, 4, and 3. Therefore, for precision coffee growing, based on this study conditions, it is recommended to use a grid from 2.0 to 1.0 points/ha, for areas above 100.0 ha and a grid with 2.0 points/ha for areas equal or smaller than 50.0 ha.

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Grid	NPG	C	C,	$C_0 + C_1$	V	SDI_Eni	oSeidel	CV(%)	$\mathbf{M}_{(SE)}$	$SD_{(SE)}$	SAI	SPI	SOGI
						Phosphor	us 2014						
] * *	52	64.92	1912.09	1977.01	112.03	16.50	Mod	39.58	0.0088	0.9526	0.4500	0.0811	26.56
2*	26	1600.37	801.01	2401.38	525.38	9.93	Weak	42.64	0.0037	1.0367	0.7688	0.0000	38.44
3*	17	1699.97	67.69	1767.66	317.38	1.38	Weak	36.57	0.0055	0.943	0.6563	0.0904	37.33
4*	13	2084.43	804.59	2889.02	684.51	11.59	Weak	38.94	0.0160	1.0191	0.0000	0.0170	0.85
						Phosphor	us 2015						
]*	224	0.00	1415.65	1415.65	130.66	14.80	Mod	42.81	0.0173	0.9120	0.4627	0.1095	28.61
2*	112	424.68	810.82	1235.50	185.00	13.75	Mod	37.64	0.0239	1.0242	0.2578	0.0000	12.89
3*	62	1159.76	0.00	1159.76	813.13	0.00	Weak	34.57	0.0138	0.9378	0.5714	0.0844	32.79
4*	57	1118.67	0.00	1118.67	821.42	0.00	Weak	36.43	0.0322	0.9670	0.0000	0.0558	2.79
						Potassiun	n 2014						
1*	52	0.00	779.92	779.92	110.60	12.53	Mod	28.63	0.0115	1.0145	0.6875	0.0305	35.90
2*	26	673.89	178.26	852.16	293.52	96.9	Weak	28.79	0.0140	1.0465	0.6195	0.0000	30.98
3*	17	816.07	69.38	885.46	813.13	3.61	Weak	27.77	0.0368	1.0224	0.0000	0.0230	1.15
4*	13	878.64	0.00	878.64	547.61	0.00	Weak	27.24	0.0165	0.9665	0.5516	0.0764	31.40
						Potassiur	n 2015						
*	52	95.11	62.41	157.52	153.23	6.88	Weak	24.95	0.0004	0.9657	0.9934	0.1074	55.04
2***	26	39.89	119.72	159.61	183.96	15.63	Mod	23.34	0.0325	0.9925	0.4646	0.0826	27.36
3*	17	75.71	29.14	104.85	470.71	7.41	Weak	19.03	0.0080	1.0819	0.8682	0.0000	43.41
4***	13	22.00	84.64	106.64	284.38	27.44	Str	19.17	0.0607	1.0781	0.0000	0 0035	0.18
NPG - Nur	nber of poin	nts of the grid	sampling; C	0 - Nugget e	effect; C ₁ -	Contributi	ion; C ₀ +C	1 - Sill; a -	range; SD	I - Spatial o	lependence	index; Str -	Strong; Mod
- Moderate	; Weak - W	⁷ eak; CV (%)	- Coefficient	t of variation	n; M _(SE) -]	Mean of st	andard er	ror; SD _{se} -	Standard	deviation o	f standard e	error; SAI -	Standardized
accuracy in	idex; SPI - S	Standardized p	precision inde	ex; SOGI - S	Standard o	ptimal grid	indicator						

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Grid	NPG	C,	C_1	$C_{0} + C_{1}$	а	SDI_S	Seidel	CV(%)	$\mathbf{M}_{(\mathrm{SE})}$	$SD_{(SE)}$	SAI	SPI	SOGI
						Calciu	um 2014						
	52	1.26	1.08	2.34	259.44	13.60	Mod	46.09	0.0064	1.0234	0.8867	0.0000	44.34
2*	26	0.79	1.18	1.98	213.01	14.45	Mod	40.75	0.0565	0.9463	0.0000	0.0753	3.77
3 *	17	1.18	0.59	1.77	284.09	10.75	Weak	38.32	0.0306	0.9467	0.4584	0.0749	26.67
4*	13	1.50	0.00	1.50	547.61	0.00	Weak	34.57	0.0237	0.9550	0.5805	0.0668	32.37
						Calciu	um 2015						
]*	100	1.66	0.89	2.55	114.87	5.07	Weak	54.04	0.0148	1.0275	0.5502	0.0000	27.51
2*	50	2.30	0.00	2.30	359.06	0.00	Weak	47.77	0.0223	0.9772	0.3222	0.0490	18.56
3*	33	2.39	0.00	2.39	505.14	0.00	Weak	49.43	0.0329	0.9982	0.0000	0.0285	1.43
4*	25	1.34	0.00	1.34	547.61	0.00	Weak	39.46	0.0187	1.0136	0.4316	0.0135	22.26
						Magnes	sium 2014						
1*	52	0.30	0.09	0.38	162.48	4.16	Weak	51.04	0.0042	1.0024	0.9302	0.0039	46.71
2*	26	0.19	0.13	0.32	244.26	11.26	Weak	45.77	0.0525	1.0063	0.1279	0.0000	6.40
3*	17	0.11	0.24	0.35	255.30	19.73	Mod	46.10	0.0602	0.9400	0.0000	0.0659	3.29
4*	13	0.26	0.08	0.34	821.42	12.04	Weak	43.69	0.0196	0.9477	0.6744	0.0582	36.63
						Magnes	sium 2015						
1*	52	0.34	0.12	0.46	120.48	3.56	Weak	61.50	0.0100	1.0399	0.5215	0.0143	26.79
2*	26	0.50	0.04	0.54	813.38	3.41	Weak	55.67	0.0065	1.0131	0.6890	0.0397	36.44
3 *	17	0.59	0.02	0.61	813.13	1.51	Weak	57.19	0.0209	1.0550	0.0000	0.0000	0.00
4*	13	0.36	0.00	0.36	821.42	0.00	Weak	46.17	0.0056	1.0240	0.7321	0.0294	38.07
NPG - Numl - Moderate; accuracy ind	ber of poin Weak - We lex; SPI - S	ts of the griv sak; CV (%) tandardized	l sampling - Coeffici precision i	;; C ₀ - Nugg lent of varia ndex; SOGI	et effect; C tion; M _(SE) - Standard	- Contrib - Mean of optimal g	oution; C ₀ + c standard e grid indicate	C ₁ - Sill; a - error; SD _{sE} or.	- range; SDI - Standard o	- Spatial d leviation of	ependence standard	index; Str - error; SAI -	Strong; Mod Standardized

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Area:	Ranking	Grid	Number of points	Mean SOGI (%)
	1 st	1	224	34.28 a
1124	1 st	2	112	27.43 a
112na	2^{nd}	4	57	20.08 b
	2^{nd}	3	79	17.32 b
	1 st	1	100	38.59 a
50h a	2^{nd}	2	50	28.78 b
Sona	2^{nd}	4	25	27.81 b
	3 rd	3	33	6.30 c
	1 st	1	52	32.58 a
2(ha	2^{nd}	2	26	22.92 b
2011a	2^{nd}	4	13	19.68 b
	2^{nd}	3	17	16.67 b

TABLE 7 - Ranking of grids as a function of the mean SOGI (%), considering soil fertility and collection seasons.

4 CONCLUSIONS

By applying the methodology, through the SAI, SPI and SOGI, it was possible to identify the most recommended grid for the tested properties.

Based on this study conditions, it is recommended the georeferenced sampling, obeying the use of grids with 2.0 points per hectare, for precision coffee growing.

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