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# BED TEMPERATURE IN COMPOST BARNS TURNED WITH ROTARY HOE AND OFFSET DISC HARROW

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#### **KEYWORDS**

aeration/oxygenation of the bed, dairy cattle, confinement, agricultural implements, spatial variability

# **ABSTRACT**

The success of confinement for dairy cattle in the compost barn model depends mainly on the management of the bed and consists of its turning. This paper characterises the spatial variability of the bed temperature in the compost barn confinement model, as well as verifying whether there was an effect on efficiency from turning the bed with different agricultural implements. The experiment was conducted during the summer and winter of 2016, with 8 days of collection in each period. Data were analysed with descriptive statistics methods and geostatistical modelling with semivariograms and kriging maps. The results of the t-test at a5% significance level indicated that, after turning the bed with the plough or rotary hoe, the temperature values were significantly lower. There was an efficient on bed turning based on different agricultural implements. Spatial dependence was observed on the data, with a better adjustment given by a Gaussian model. Kriging maps allowed the characterization of the spatial variability of bed temperature and the visualization across the compost barn bed at the superficial layer and 0.15 m in depth.

# INTRODUCTION

The confinement of dairy cattle in a compost barn (CB) is an alternative system of the Loose Housing system, where the animals stay loose and can walk freely inside a shed (rest area), influencing the welfare of the animals and, consequently, improving herd productivity levels (Black et al., 2013, Mota et al., 2018).

The success of the system depends mainly on the proper handling of the bed, which consists of its turning (Mota et al., 2017, Pilatti & Vieira, 2017). The management of bedding material provides a dry, comfortable and healthy environment in which cows can stand and walk on a soft surface (Leso et al., 2013). The most common materials used as bedding are sawdust and wood chips (Pilatti & Vieira, 2017). Small particulate materials such as finely processed straw, corn straw and wheat straw by-products can also be used. However, care should be taken with other types of bedding materials, such as sand, dry manure or soil (Galama et al., 2015).

The bed needs to be turned frequently and this can be observed in CB experiments for dairy cows reported in the literature of USA, Israel and Italy, among others (Leso et al., 2013, Black et al., 2014). The indicated average temperature should be between 54.4 °C and 65.5 °C; the ideal bed humidity should range from 40 and 60% (Black et al., 2013), the C: N ratio should be 25–30:1 and the

recommended animal density is 7.4–12.5 m<sup>2</sup> cow<sup>-1</sup> (Galama et al., 2015). These values should be controlled because the entire surface of the resting area is covered with a deep bed material (soft bed), which is often agitated by agricultural implements to incorporate fresh manure into that material and increase water evaporation (Leso et al., 2013; Galama et al., 2015).

This process of bed turning is necessary for aeration to occur, and this maintains its aerobic condition. The process is usually carried out when the cows go to the milking parlour. Normally the agricultural implements used for bed turning are rotary hoes, subsoilers and offset disc harrows (Galama et al., 2015, Mota et al., 2017).

The harrow revolves the bed through the discs that are widely spaced, and when cutting the bedding material, the disc plough removes semi-circular sections that help in the incorporation of oxygen to that material. The subsoiler is used to turn the deeper layers of the bed, preventing the deeper regions from becoming anaerobic, while the rotating hoe does not turn the deeper parts of the bed, but breaks up the materials that become compacted. In this manner, this decomposition of aggregate particles promotes oxygenation and increases the surface action of the microorganisms present in the bed, causing an increase in composting efficiency (Mota et al., 2017).

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The composting bed CB refers to a mixture of faeces and urine produced by cows and organic bedding material (Leso et al., 2013). The biological activity generates heat and helps to dry the bedding material, and when this material begins to adhere to the cows, a clean material should be added to the bed (Galama et al., 2015).

Despite the increasing popularity of the CB system in Brazil, the scientific knowledge about this system is scarce and consequently requires more studies that can assist the milk producers in the process of decision making and planning. Therefore, it is important to encourage research related to dairy cattle facilities to maintain production at competitive levels. The objective of this research was to characterise the spatial variability of the bed temperature in the confinement system in the compost barn model, as well as to verify if there was an effect on bed turning efficiency based on different agricultural implements.

## **MATERIAL AND METHODS**

The research was carried out in a compost barn confinement for dairy cattle, on a rural property in the municipality of Três Corações in the state of Minas Gerais. According to the Köppen international classification, the climate of the region is of the Cwa type, characterised by two well-defined seasons: a dry and cooler temperature, which extends from May to September, and a humid and warmer temperature from October to April. The average annual temperature is 20.2 °C, and the annual rainfall is 1.401 mm.

The experiment was conducted during the summer (VER treatment) and winter (INV treatment) of 2016 and was divided into 8 days of collection each for the VER and the INV. The data recording was carried out on January 10, 13, 17, 20, 24, 27, 31 and February 3 for summer and on July 10, 13, 17, 20, 24, 27, 31 and August3 for winter.

Data were recorded using a model 3000 Kestrel® brand portable data recorder that measured relative air humidity with  $\pm$  3% reading accuracy and temperature (accuracy  $\pm$  1 °C). The average air temperature and the external relative humidity of the shed during the data collection days were obtained with a model TTWH–1080 Instrutemp portable weather station with appropriate sensors to collect and store the data.

The shed in system compost barn (CB) with a Northwest to Southwest orientation, was 18.70 m wide by 50 m long, with a 4 m right foot, and a 6 m ridge. A  $13 \times 50$  m rest area with beds was divided into 3 lots (high milk production, average milk production and low milk production), separated by means of electric fencing, and this was the area used for analysis. The shed had a feeding corridor measuring  $3 \times 50$  m with a feeder (trough of bulk) throughout the length of the shed. There was another corridor measuring  $2 \times 50$  m with four drinking troughs measuring  $0.50 \times 2$  m (trapezoidal bottoms) with the capacity for 300 L of water. The shed also had three mineral troughs and five centralised fans at a height of 3.20 m in the Roster seating area (with six propellers). Each fan provided a flow of 48,000 m $^3$  of air per hour.

The superficial temperature data (S) of the CB bed were collected using a model GM-300 laser sight IR thermometer, with a temperature range of -50–380 °C (-50 at 0 °C) +/- 2 °C and (0 to 380 °C) +/- 1.5 °C. The bed temperature data for the CB at a depth of 0.15 m (P) were collected by a model ICEL Manaus TD-100 thermometer. There were 36 systematic samples collected for S and P, before and after to be the bed was turned during the afternoon period, at the time of the second milking at 16:00 hours, in a regular grid (Yamamoto & Landim, 2013) with 36 georeferenced points, distributed evenly along the shed with a  $2.6 \times 5$  m spacing (Figure1).

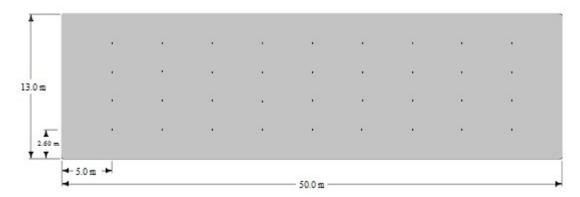


FIGURE1. Regular grid with 36 points used to record the bed superficial (S) and 0.15 m depth (P) temperatures.

The agricultural implements evaluated were a Santa Isabel model harrow with 28 discs angled of approximately  $20^{\circ}$  that were set to a 0.20 m depth and a Selecta tilth model rotary hoe with seven cutting knives, with a depth adjustment of 0.0–0.20m. There were eight days of sample collections in the summer of 2016, using the harrow for turning the bed (Figure 2a) and eight days of collection in the winter of 2016 using the rotary hoe (Figure 2 b).





a

FIGURE2. Agricultural implements used to turn the bed. (a) disc harrow; (b) rotary hoe.

Initially, data for all collection days and the data average for the eight days each of summer and winter collection were analysed using descriptive statistics analysis procedures. The objective was to visualise the general behaviour of the data, determining position and dispersion measures, such as the mean, median, maximum (max) and minimum (min) values, standard deviation and coefficient of variation (Silva et al., 2012). Then the Shapiro Wilk test was applied to verify the normality of the data, the F test to compare two variances and the means test to compare paired data (Torman et al., 2012, Oliveira et al., 2014). There was a level of significance of  $\alpha > 0.05$  for all of the tests mentioned. Subsequently, only the data referring to the average of the eight days of summer collection and the average of the eight days of winter collection were submitted to geostatistical modelling to verify the spatial variability of the data. This was done through the construction of semivariograms and kriging maps.

The semivariograms were estimated using the classical Matheron estimators, given by (Yamamoto & Landim, 2013):

$$\hat{\gamma}(h) = \frac{1}{2N(h)} \sum_{j=1}^{N(h)} [z(x_j + h) - z(x_j)]^2$$
 (1)

Where;

N(h) is the number of possible pairs for the distance h;

 $\hat{\gamma}(h)$  is the semivariance for a distance h;

 $z(x_j)$  e  $z(x_j + h)$  are the observations, surface temperature and temperature at 0.15 m separated by the vector h, and

h is the separation distance of the observations.

For the adjustment of the theoretical semivariogram, the OLS method was used. The spherical, Gaussian, exponential and linear isotropic statistical models were compared according to Yamamoto & Landim (2013), and they are defined by:

Spherical Model

$$\gamma(h) = \begin{cases} C_0 + C \left[ \frac{3}{2} \left( \frac{h}{a} \right) - \frac{1}{2} \left( \frac{h}{a} \right)^3 \right] & 0 \le h \le a \\ C_0 + Ch > a \end{cases}$$
 (2)

Gaussian Model

$$\gamma(h) = C_0 + C \left( 1 - exp\left( -3\left(\frac{h^2}{a^2}\right) \right) \right) \text{ , se } 0 \le h \le d \quad (3)$$

**Exponential Model** 

$$\gamma(h) = \begin{cases} C_0 + C \left[1 - e^{\left[-\left(\frac{h}{a}\right)\right]}\right] & \text{para } 0 < h < a \\ C_0 + C & \text{para } h > a \\ 0 & \text{para } h = a \end{cases}$$
 (4)

Linear Model

$$y(h) = \begin{cases} C_0 + \frac{c}{ah} & 0 \le h \le a \\ C_0 + C & h > a \end{cases}$$
 (5)

Where;

 $\gamma(h)$  is the semivariance for the distance h;

 $C_0$  is the nugget effect;

 $C_0 + C$  is the sill;

a is the range of spatial dependence, and

 $\frac{c}{a}$  is the angular coefficient for  $0 \le h \le a$ .

From the set of results obtained by the classic estimators for the three models, their parameters were estimated, and the best model was selected according to the approximation of the value of the Akaike Criterion (AIC), given by:

$$|AIC| = 2p + n \cdot ln\left(\frac{RSS}{n}\right),$$
 (6)

Where;

*n* is the number of observations;

p is the number of parameters, and

RSS is the sum of squares of residues, and the model with the lowest value of AIC was considered to be the best (Mota et al., 2008).

The relationship:

$$\left(\frac{co}{co+c_1}\right) x 100, \tag{7}$$

was used to evaluate the degree of spatial dependence (SD) of the variables the according to Cambardella et al. (1994), which classifies strong, moderate and weak SD

values, when they have a nugget effect < 25%, 25-75%, and > 75% of the sill, respectively. If the relation is greater than 100%, the variable is considered spatially independent.

When spatial dependence occurs, the values not measured can be estimated without trends and with minimum variance in order to obtain information about the variable (through mapping). In these cases, it is necessary to use a geostatistical interpolator to generate a smoothed surface of the contour maps. Among the several interpolators in the literature, ordinary kriging has been used, which estimates everywhere, except where field observations are available, at which it reproduces the measured value (Yamamoto & Landim, 2013). To verify if ordinary kriging adequately described the spatial variability of the variables studied, cross validation was used.

R<sup>2</sup> (coefficient of determination), and the RSS (sum of squares of residues) were used as criteria for the cross-validation. The best fit was reached when the correlation and determination coefficients were close to a value of 1, and there was an adjacent zero intercept and an adjacent angular coefficient of 1 (Assumpção & Hadlich, 2017). For RSS, smaller values correspond with better semivariogram models. The analyses were performed in R software (R Core Team, 2016) and GS+.

#### RESULTS AND DISCUSSION

The mean air temperature and the average relative air humidity inside the shed during the summer and winter collection days are shown in Figures 3 and 4.The mean values of external temperature in the summer and winter of 2016 were 26.5 and 21.2 °C, respectively, and mean humidity values were 72% in summer and 58% in winter.

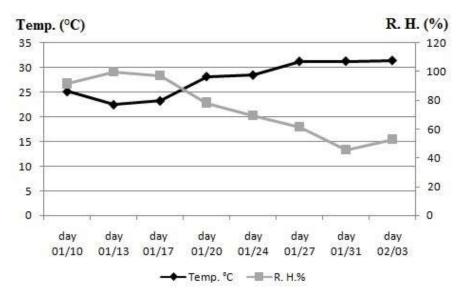


FIGURE 3. Variation of temperature (°C) and relative humidity (%) inside the experimental shed in the summer period.

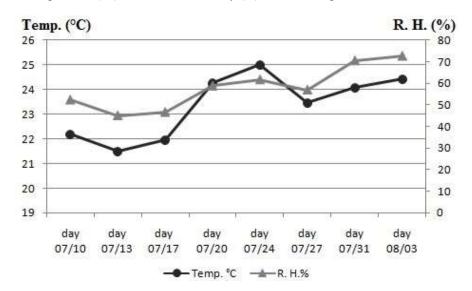


FIGURE 4. Variation of temperature (°C) and relative humidity (%) inside the experimental shed during the winter period.

From the average temperature of the confinement bed in the compost barn model (Table 1), the maximum temperature at a depth of 0.15 m (P) obtained in summer 2016 before being turned was around 40.9 °C and after turning around 41.8 °C. This was below the ideal for this model of confinement. The average temperature indicated

according to Leso et al. (2013) and Black et al. (2014) must be between 54.4 the 65.5 °C. The same was observed with the maximum temperature at the depth of 0.15m (P), obtained in winter 2016, although it had higher values, 42.8 and 44.3 °C, before and after turning, respectively.

TABLE 1. Descriptive statistics and paired t-test results ( $\alpha = 0.05$ ) for the mean values of the eight days of superficial (S) and depth of 0.15 m (P) temperature collection of bedding in the compost barn.

	(G - V	(G - VER)		(E -INV)	
	Before	After	Before	After	
Min. (S)	22.912	24.525	15.762	17.337	
Min. (P)	26.572	29.557	25.044	30.200	
Max. (S)	26.543	29.481	22.050	22.762	
Max. (P)	40.894	41.831	42.722	44.275	
Median (S)	24.387	26.140	17.925	19.250	
Median (P)	30.922	33.584	29.433	33.231	
Mean(S)	24.537	26.139	18.167	19.525	
Mean (P)	32.117	34.054	30.746	34.343	
Standard deviation (S)	0.864	1.277	1.401	1.301	
Standard deviation (P)	3.554	2.677	4.230	3.555	
Coefficient of Variation (S)	3.52%	4.88%	7.71%	6.66%	
Coefficient of Variation (P)	11.06%	7.86%	13.75%	10.35%	
Paired t-test (S)	5.817	5.817e <sup>-11*</sup>		5.775e <sup>-06*</sup>	
Paired t-test (P)	2.455	2.455e <sup>-14</sup> *		1.218e <sup>-13*</sup>	

<sup>\*</sup> significant at a 5% probability. Harrow in summer (G - VER) and rotating hoe in winter (E - INV).

The surface temperature was within the expected range, with higher values in summer and lower temperatures in winter, for both minimum (Min) and maximum (Max) temperature values before and after being turned, using the harrow in summer and the rotating hoe in winter. The coefficient of variation is a way of expressing the variability of the data, so when the value of the coefficient of variation was smaller the dispersion around the mean was smaller. The superficial and 0.15m temperatures before and after the bed was turned had low coefficients of variation ( $CV \leq 15\%$ ), indicating data homogeneity and low dispersion (Nazareno et al., 2016).

The data were normal, and the variances were homogeneous according to the F test for the VER treatment, so the t-test was applied to the paired samples. The results indicated that there was a significant difference between the means of the two samples, and we concluded that the differences in CB bed temperature before and after being rotated with the harrow were significant.

For the rotational hoe used in the winter, the data also had a normal distribution according to the Shapiro-Wilk normality test, and the variances were homogeneous, so the paired t-test was used.

The results of the t-test at the significance level of 5% indicated that, after turning the bed with the harrow or rotating hoe, the values of the temperatures were significantly lower, indicating that there was an efficiency effect on the bed turning from these agricultural implements.

The bedding material needs to be turned daily for aeration to occur and to maintain aerobic conditions. Biological activity produces heat that assists in drying the bed, and when this material begins to adhere to cows, a clean, dry material should be added (Galama et al., 2015).

Mota et al. (2018), states that turning the material of the bed properly reduces the moisture and increases bed temperature, improving the composting process, with reductions in pathogenic microorganisms.

From the results of the geostatistical modelling in summer (for the theoretical models and estimated parameters of the semivariograms), a strong spatial dependence was observed for temperature at the surface and at the depth of 0.15m of the confinement bed for dairy cattle in the compost barn model in the summer of 2016 (Table 2). The exception was the linear model that showed a weak spatial dependence for the surface temperature before being turned and a moderate dependence for the other variables.

For the superficial temperature in summer, there was a better fit for the "Gaussian" model, with lower AIC and RSS values when compared with the spherical, exponential and linear models. As for temperature at the depth of 0.15 m during the summer of 2016, the results were similar. A better adjustment of the "Gaussian" model was observed, with a high degree of spatial dependence, or the semivariograms had a nugget effect equal to or less than 25% of the level and with lower AIC and RSS values while having higher R² values.

TABLE 2. Estimation of the nugget effect parameters  $(C_0)$ , sill  $(C_0 + C_1)$ , range (a), approximation of value of the Akaike Criterion (AIC), degree of spatial dependence (SD), coefficient of determination  $(R^2)$  and the sum of squares of residues (RSS) of the "spherical" (Sph.), "Gaussian" (Gaus.), "exponential" (Exp.) and "linear" (Lin.) adjusted to the experimental semivariograms, relative to the superficial and 0.15 m depth temperatures of the confinement bed for dairy cattle in the compost barn model in the summer of 2016.

1										
	Superficial - summer									
	Before				After					
	Sph.	Gaus.	Exp.	Lin.	Sph.	Gaus.	Exp.	Lin.		
a	6.69	5.438	7.74	17.0	6.84	5.96	9.87	17.0		
$C_0$	0.001	0.001	0.001	0.66	0.001	0.141	0.001	1.41		
$C_0 + C_1$	0.824	0.829	0.837	0.846	1.946	1.965	2.051	2.09		
AIC	-9.37	-9.46	-8.81	-7.83	-1.69	-1.73	-1.64	-1.06		
SD%	0.121	0.120	0.119	78.014	0.051	7.175	0.048	67.464		
$\mathbb{R}^2$	0.453	0.454	0.342	0.054	0.259	0.263	0.243	0.073		
RSS	0.214	0.208	0.258	0.358	2.77	2.73	2.82	3.42		
	Depth 0.15m - summer									
	Before				After					
	Sph.	Gaus.	Exp.	Lin.	Sph.	Gaus.	Exp.	Lin.		
a	8.51	4.12	3.74	17.0	8.14	3.99	4.02	17.0		
$C_0$	0.01	0.01	0.01	9.45	0.01	0.01	0.01	5.18		
$C_0 + C_1$	14.32	14.55	14.97	14.99	8.71	8.85	9.33	9.65		
AIC	11.375	11.174	11.628	12.380	7.177	6.918	7.441	8.284		
SD%	0.069	0.068	0.066	63.042	0.114	0.112	0.107	53.678		
$\mathbb{R}^2$	0.366	0.386	0.302	0.057	0.45	0.468	0.375	0.13		
RSS	216	202	235	302	53.3	48.9	58.2	77.1		

The values of AIC and SD and the estimates of the parameters of the theoretical semivariogram models for the surface temperature and temperature in the depth of 0.15m of the confinement bed for dairy cattle in the compost barn model in the winter of 2016 are shown in Table 3.

A better fit of the "Gaussian" model was observed, with moderate spatial dependence on before and after surface temperature and a strong degree of spatial dependence for the temperature at a depth of 0.15m.

TABLE 3. Estimation of the nugget effect parameters  $(C_0)$ , sill  $(C_0 + C_1)$ , range (a), approximation of value of the Akaike Criterion (AIC), degree of spatial dependence (SD), coefficient of determination (R<sup>2</sup>) and the sum of squares of residues (RSS) of the "spherical" (Sph.), "Gaussian" (Gaus.), "exponential" (Exp.) and "linear" (Lin.) models adjusted to the experimental semivariograms, relative to the superficial temperature and temperature in the 0.15 m depth of the confinement bed for dairy cattle in the compost barn model in the winter of 2016.

				Super	ficial - winter				
		Before				After			
	Sph.	Gaus.	Exp.	Lin.	Sph.	Gaus.	Exp.	Lin.	
а	33.60	16.49	20.15	17.0	30.71	15.63	23.78	17.00	
$C_0$	0.552	0,921	0.41	0.591	0.369	0.558	0.314	0.3969	
$C_0 + C_1$	3.24	3.32	3.83	2.47	1.642	1.698	2.25	1.34	
AIC	- 3.30	-3.37	-3.21	-3.34	-7.31	-7.34	-7.26	-7.34	
SD%	17.04	27.74	10.70	23.93	22.47	32.86	13.95	29.61	
$\mathbb{R}^2$	0.56	0.57	0.55	0.56	0.553	0.559	0.547	0.558	
RSS	1.62	1.58	1.67	1.60	0.426	0.421	0.432	0.421	
	Depth 0.15 m - winter								
	Before				After				
	Sph.	Gaus.	Exp.	Lin.	Sph.	Gaus.	Exp.	Lin.	
а	10.27	8.07	13.83	17.00	8.07	6.72	10.83	17.00	
$C_{0}$	0.01	0.01	0.01	11.84	0.01	0.01	0.01	9.69	
$C_0 + C_1$	23.81	23.91	25.26	26.22	15.31	15.49	16.02	16.65	
AIC	10.94	10,44	11.68	13.08	6.69	6.01	7.781	9.56	
SD%	0.042	0.042	0.039	45.16	0.065	0.064	0.062	58.19	
$\mathbb{R}^2$	0.67	0.70	0.57	0.24	0.76	0.78	0.60	0.19	
RSS	187.0	158.0	239.0	381.0	45.3	36.1	65.2	118.0	

The results obtained from the analysis of bed temperature behaviour corroborate with those presented by Carvalho et al. (2011) in his research. These authors analysed the quality of reused chicken beds and air in commercial production aviaries with different ventilation systems and typology through geostatistical analysis. They identified spatial dependence, characterised by adjustment of the theoretical models of semivariograms, and with better adjustment of the spherical model to the bed temperature of the aviary, the authors concluded that the

geostatistical analysis assisted in the identification of critical points in the control of the environments studied.

Figure 5 and 6 show the bed superficial temperature distribution and the temperature at 0.15 m depth for the VER and INV treatments, using the harrow and the rotary hoe, respectively. In the surface temperature of the bed in the treatment summer (Figure 5) before and after being turned by the grid harvester, it was observed that there was a small change, with higher values in proximity to the electric fence that separates the pickets with small islands of temperature above 25.6 °C.

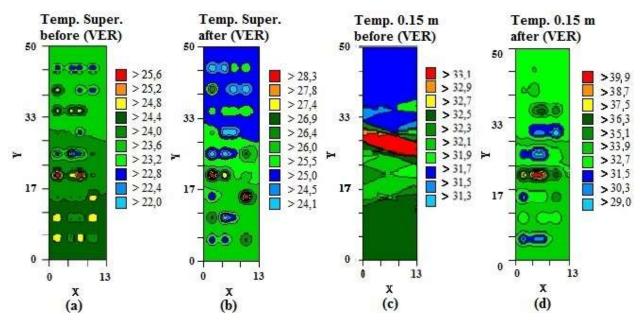


FIGURE 5. Maps of kriging for summer 2016. (a) surface temperature of the bed before being stirred, (b) bed surface temperature after turning, (c) bed 0.15m temperature before turning and (d) 0.15 m temperature of the bed after it was turned.

For the temperatures at 0.15 m depths, lower temperatures were observed in the region near the exit of the shed, where the afternoon sun was not predominant, which indicated that the position of the installation influenced the homogeneity of the bed temperature. After

turning, with the stirring of the bed material, there was a greater homogeneity of temperature throughout the shed, maintaining the highest temperature near the picket divisions, with temperatures above 39.9 °C.

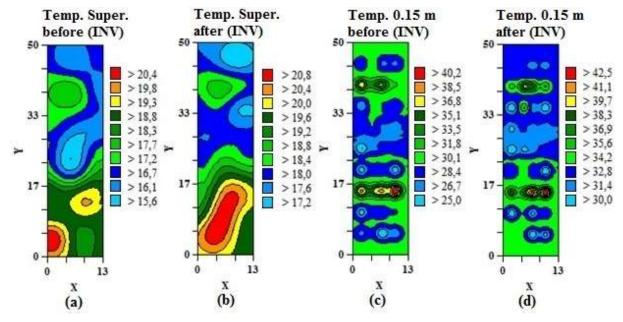


FIGURE 6. Maps of winter kriging of 2016.(a) surface temperature of the bed before being stirred (b), bed surface temperature after turning (c) 0.15 m bed temperature before turning and (d) 0.15 m temperature of the bed after it was turned.

In winter (Figure 6), the surface temperature had higher values near the entrance with islands (red dots on the map) of 20.4 °C, which indicated that the afternoon sun influenced this temperature. Additionally, in the other part of the shed the afternoon surplus and dominant surface temperatures were higher than 15.6 °C, and, even after turning with the rotating hoe, the difference at that point was small, increasing to 17.2 °C. While in the winter, the temperatures at 0.15 m had presented higher values (42.5 °C) when compared to the summer, with islands of temperature also in the divisions of the pickets, which are the divisions with the electric fence (the colours red and orange in the map).

## **CONCLUSIONS**

Geostatistics allowed us to characterise the variability and spatial dependence of the surface and 0.15 m depth temperatures of the bed in the compost barn confinement during the summer and winter seasons, for milk cattle. The turning of the bed by means of the harrow and rotary hoe enabled the reduction of temperature in both assessment layers.

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