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Mathematical models of anaerobic digestion for the treatment of swine effluents¹

Daisa Stéfano Fagundes², Marco Antonio Previdelli Orrico Junior², Ana Carolina Amorim Orrico², Leonardo Oliveira Seno²

RESUMO

Modelos matemáticos da biodigestão anaeróbia para o tratamento de efluentes de suínos

A suinocultura resulta na produção de grandes quantidades de resíduos, os quais, quando manuseados incorretamente, podem causar graves problemas ambientais. Este estudo objetivou testar modelos matemáticos para estimar os parâmetros de biodigestão anaeróbia em biodigestores, em função da variação da composição de efluentes de suinocultura com e sem separação da fração sólida e dos tempos de retenção hidráulica (TRH). Foram utilizados biodigestores semicontínuos, abastecidos com efluentes de suinocultura com e sem separação da fração sólida, manejados por 15, 22, 29 e 36 dias de retenção hidráulica. Foram avaliados os potenciais de produção de biogás e metano e as reduções de sólidos totais, sólidos voláteis e demandas químicas de oxigênio, em função da composição dos efluentes. O TRH foi a variável que teve maior influência na variação dos modelos, seguido dos teores de sólidos totais e voláteis. Os modelos uni e multivariados apresentaram elevados índices de confiança, sendo classificados como ótimos para predizer os potenciais de produção de biogás e metano e como bons para predizer as reduções de sólidos totais, sólidos voláteis e demanda química de oxigênio. Os modelos deste estudo podem ser utilizados para predizer com segurança os parâmetros do processo de biodigestão anaeróbia de efluentes de suinocultura, em biodigestores semicontínuos tubulares.

PALAVRAS-CHAVE: Metano; biodigestor; biogás.

INTRODUCTION

The Brazilian pork industry, like other Brazilian agribusiness production chains, has grown significantly in recent years (ABIPECS 2013). However, it is considered an activity with high potential for pollution, because of the large volume of effluents produced per area.

Swine manure effluents are composed of feces, urine, food waste, spilled water, hair and dust from the raising process (Orrico Junior et al.

ABSTRACT

Pig breeding results in the production of large amounts of waste, which can cause serious environmental problems, when handled incorrectly. This study aimed at testing mathematical models to estimate the parameters of anaerobic biodigestion in biodigesters as a function of the composition of swine effluents with and without separation of the solid fraction and hydraulic retention times (HRT). Semi-continuous biodigesters fed with swine effluents with and without separation of the solid fraction and managed for 15, 22, 29 and 36 days of hydraulic retention were used. The potential of biogas and methane production, as well as the reduction of total solids, volatile solids and chemical oxygen demand, were assessed as a function of the effluents composition. HRT was the variable that most influenced the variation of the models, followed by the contents of total and volatile solids. Uni and multivariate models presented high confidence indices, being classified as "great" at predicting the potentials of biogas and methane production and "good" at predicting the reductions of total solids, volatile solids and chemical oxygen demand. The models obtained in this study can be used to reliably predict the parameters of the anaerobic biodigestion process of swine effluents in semi-continuous tubular biodigesters.

KEY-WORDS: Methane; biodigester; biogas.

2010). When effluents are mishandled, chemical and microbiological contamination of soil and consequently of surface and groundwater may occur, in addition to air pollution by emissions of greenhouse gases (Dinuccio et al. 2008, Babson et al. 2013).

The anaerobic digestion process by implementation of digesters can be considered an interesting process, because it promotes waste treatment and generates renewable energy that can return to the production system by converting chemical energy of biogas into electrical or thermal energy (Flesch et al. 2011).

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The potential of biogas and methane production depends mainly on the composition of the effluent (contents of essential nutrients and solids) and the time in which the material remains in the digester (hydraulic retention time - HRT). In general, the greater the amount of nutrients, the greater the HRT required to stabilize the effluent.

Sieving the effluent is a way to improve the efficiency of the process, since it separates the highly biodegradable liquid fraction from the solid fraction, which has slower degradation (Orrico Junior. et al. 2010). Despite these benefits in the anaerobic digestion process, many producers do not separate the effluents, because the solid fraction retained on the sieve must undergo additional slower treatment steps, such as composting, what often discourages the producer (Orrico Junior et al. 2009).

In the literature, most of the studies only quantify the volume of biogas produced by digesters supplied with swine effluent without considering the composition of the effluent and the HRT. The composition and digestion time affect biogas production (Herrero 2011), therefore, the development of mathematical models that take these factors into account to estimate the main parameters of the anaerobic digestion process can help to design digesters and predict their performance (Astals et al. 2013).

The present study aimed to test mathematical models to estimate the parameters of anaerobic digestion on biodigesters as a function of the variation of the composition of effluents from the pork industry, with and without separation of the solid fraction, and according to the hydraulic retention times (HRT).

MATERIAL AND METHODS

The experiment was conducted between March and September 2013, in Dourados, Mato Grosso do Sul State, Brazil, in a completely randomized design, following a 2 x 4 factorial scheme, with four replicates per treatment.

Data were obtained from semi-continuous tubular digesters fed with swine effluent with and without sieving, managed by 15, 22, 29 and 36 days of hydraulic retention. Some properties and nutrient contents of the swine effluent used are shown on Table 1.

The laboratorial digesters consisted of two distinct parts: a container that holds the material for

Table 1. Distribution of minimum (Min), maximum (Max), mean value (MV) and standard deviation (SD) of components of swine effluents (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Variable	Min	Max	MV	SD
TS (%)	1.23	2.20	1.74	0.38
N (% in TS)	3.03	4.41	3.77	0.34
P (% in TS)	3.25	5.64	4.14	0.56
VS (%)	1.09	1.98	1.21	0.68
K (% in TS)	4.72	9.17	7.07	1.39
CH ₄ (%)	63.10	75.90	69.65	4.27
CH ₄ VSadd (L/ kg VSadd)	448.12	1,160.00	760.47	250.12
TSR (%)	47.24	76.19	63.56	6.76
VSR (%)	50.44	80.04	68.85	6.89
CODR (%)	57.66	82.55	69.19	7.75

TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium; CH₄ = % methane; CH₄ VSadd = methane production per kg of added VS; TSR = total solid reduction; VSR = volatile solid reduction; CODR = chemical oxygen demand reduction

fermentation (volume of 0.04 m³) and a gasometer (Figure 1). The container was built with a PVC cylinder (300 mm of diameter and 1,000 mm in length), with the two ends fixed with PVC plates of 15 mm thick. The inlet pipe for supplying the swine effluent was fixed on one of the plates and, on the other end, two pipes were connected, one for the output of biofertilizers and another for gas exhaustion. The gasometer was composed of two cylinders of 250 mm and 300 mm in diameter, which were inserted into each other, so that the space between the outer and inner cylinders would hold a volume of water ("water seal") with a depth of 500 mm. The cylinder of 300 mm in diameter was fixed to a 25 mm thick PVC board. The 250 mm diameter cylinder had one end sealed with a cap that received the gas generated and the other end was overturned in the water seal to store the produced gas. The gasometers were placed on a bench under room temperature conditions and protected from sunlight and rain.

Biogas composition was determined using a Finigan GC-2001 gas chromatograph equipped with Porapack Q columns and molecular sieve and a thermal conductivity detector. The amount of total solids (TS), volatile solids (VS) and chemical oxygen demand (COD) were determined as described by APHA (2005).

The biogas and methane production was assessed, as well as the reductions in total solids, volatile solids and chemical oxygen demand. These data were analyzed with the R software (version 2.15.2 for Windows). The Shapiro Wilk test was used

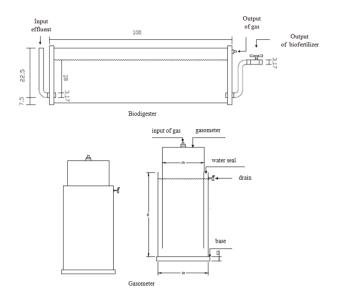


Figure 1. Scheme of semi-continuous digesters and gasometers.

to verify the normality of residuals and the Bartlett test to verify the homogeneity of variances. The presence of outliers was also evaluated. All traits studied satisfied the assumptions of the model.

After preliminary analysis, multiple regression analysis between the levels of the effluent composition (independent variables) and the characteristics of biogas production (dependent variables) were performed. The stepwise procedure of the stat package (R Development Core Team 2012) was used to indicate the independent variables that accounted for a large proportion of variability in the biogas production and composition. The selection of models was based on the coefficient of determination

(R²), root mean square error (RMSE) and Akaike information criterion (AIC).

The quality of the proposed models was assessed by using the concordance index "d" (Willmott et al. 1985) and the confidence index "c" (Camargo & Sentelhas 1997). The "c" index allows interpretation for tests of accuracy, represented by the "d" index. Precision is represented by the correlation coefficient, which is calculated from the ratio of these two indices and ranges from great (> 0.85) to very good (0.76-0.85), good (0.66-0.75), medium (0.61-0.65), fair (0.51-0.60), poor (0.41-0.50) and very poor (≤ 0.40).

The Spearman correlation (Agricolae package, version 1.1-4) was estimated to assess the association between the levels of the effluent composition (independent variables) and the characteristics of biogas production, with a significance level of 95 %. Through correlation analysis it was possible to identify the variables that most influenced the process of anaerobic digestion which were selected to build the prediction models.

RESULTS AND DISCUSSION

The production of CH₄ and CH₄VSadd were positively correlated (0.45 and 0.47) with HRT (i.e., the higher retention time of the material in the digester was essential to maximize the production of biogas and methane) (Table 2). The same was observed for TSR (0.39), VSR (0.40) and CODR (0.71).

Orrico Junior et al. (2010), using digesters fed with wastes from fully grown pigs, observed greater

Table 2. Spearman correlation among components of swine waste and biogas production, potential of biogas and methane production and reductions of total solids, volatile solids and chemical oxygen demand (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Variable	HRT	TS	N	P	VS	K	CH ₄	CH ₄ sva	TSR	VSR	CODR
HRT	-	ns	ns	ns	ns	ns	0.45**	0.47**	0.39*	0.40*	0.71***
TS	-	-	0.65***	0.67***	0.92***	-0.42*	-0.77***	-0.75***	0.56***	0.37*	0.56***
N	-	-	-	0.52**	0.77***	ns	-0.60***	0.55**	ns	ns	ns
P	-	-	-	-	0.79***	ns	-0.79***	-0.80***	ns	ns	ns
VS	-	-	-	-	-	ns	-0.78***	-0.75***	0.49*	ns	0.50**
K	-	-	-	-	-	-	0.36*	0.40*	ns	ns	ns
CH_4	-	-	-	-	-	-	-	0.98***	ns	ns	ns
CH ₄ sva	-	-	-	-	-	-	-	-	ns	ns	ns
TSR	-	-	-	-	-	-	-	-	-	0.96***	0.61**
VSR	-	-	-	-	-	-	-	-	-	-	0.48**
CODR	-	-	-	-	-	-	-	-	-	-	-

HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium; CH_4 = % methane; CH_4 sva = added volatile solids methane; TSR = total solids reduction; VSR = volatile solids reduction; CODR = chemical oxygen demand reduction; ns = non-significant; * p < 0.05; ** p < 0.01; *** p < 0.001.

reductions of TS (29 % and 44 %), VS (31 % and 45 %), COD (45 % and 61 %) and BOD (45 % and 60 %) with increasing HRT (30, 60, 90 and 120 days). The $\mathrm{CH_4}$ production also increased as manure remained longer in the digester (39 % and 82 %).

Durand et al. (1986) observed high correlation between the results obtained from experimental analyses and those generated from prediction models, with only 6 % of divergence for methane production in continuous-feed digesters with 15 days of hydraulic retention. The authors also reported that the methane production peaks varied in 6 hours between observed and estimated results, while the maximum rate of VS was 4 % superior on the results of the MI prediction model.

The analysis of univariate models indicated that the HRT and P had a significant effect (p = 0.05) on methane (Table 3). The results of R^2 , RMSE and AIC were respectively 0.91, 0.59 and 12.74 for HRT and 0.34, 57.11 and 133.44 for P. The univariate model that showed the best performance in the estimation of methane content was HRT, with a R^2 of 0.91. The multivariate linear regression obtained through stepwise (CH₄ = 0.91 * HRT + 0.02 * VS + 0.02 * TS) had a higher R^2 (0.94), highlighting HRT, TS and VS as the best variables to predict methane content. The only variable that contributed to the high

 R^2 was HRT, explaining the changes of the methane (CH_4) levels.

Methanogenesis is the final stage of the anaerobic digestion process, so the higher HRT results in higher CH₄ content in the biogas. Fdez-Güelfo et al. (2011) studied the influence of particle size and organic matter content of municipal solid waste, using semi-continuous digester with two types of composting residues (synthetic and industrial), and observed that the longer the effluent remains in the digester the greater is the production of methane.

As stated by Durand et al. (1986), the highest correlations of methane production from swine manure in continuous-feed digesters occurred with HRT, organic loading rate (concentration of volatile solids) and temperature. However, according to these authors, the inclusion of volatile solids was most correlated (0.87) to methane concentration in the biogas, when compared with TRH and temperature, and the most beneficial levels of VS for the process ranged between 60 g and 96 g VS per liter of substrate, a condition that also enhanced microbial growth.

The significant univariate models for the potential production of methane (CH₄VSadd) were the HRT, N and P. HRT showed high R² (0.89) and lower RMSE (23.68) and AIC (105.27), when compared to N and P.

Table 3. Univariate and multivariate models to estimate methane content and potential of methane production of swine effluents with no solid fraction separation (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Model	R ²	RMSE	AIC	р	Index d	r	Index c	Performance
Univariate								
$CH_A = 59.80 + 0.24 * HRT$	0.91	0.59	12.74	< 0.001	0.98	0.96	0.93	Great
$CH_4 = 67.10 - 0.62 * TS$	0.00	2.01	24.39	0.121	0.06	0.05	0.00	Very poor
$CH_{A} = 68.36 - 0.64 * N$	0.01	2.00	24.39	0.151	0.10	0.09	0.01	Very poor
$CH_4 = 77.69 - 2.61 * P$	0.35	1.62	19.44	0.010	0.70	0.59	0.42	Poor
$CH_4 = 67.10 - 0.69 * VS$	0.00	2.01	26.35	0.232	0.06	0.05	0.00	Very poor
$CH_4 = 68.91 - 0.48 * K$	0.06	1.95	25.37	0,213	0.35	0.25	0.09	Very poor
Multivariate								
$CH_4 = 57.62 + 0.24 * HRT + 7590.12 * TS - 8443.02 * VS$	0.94	0.50	23.80	< 0.001	0.98	0.97	0.95	Great
Univariate								
$CH_4VSadd = 327.81 + 8.46 * HRT$	0.89	23.68	105.27	< 0.001	0.97	0.94	0.91	Great
$CH_4VSadd = 556.02 - 6.06 * TS$	0.00	70.30	138.09	0.198	0.02	0.01	0.00	Very poor
$CH_4VSadd = 582.54 - 9.87 * N$	0.00	70.25	138.09	0.291	0.01	0.04	0.00	Very poor
$CH_4VSadd = 951.40 - 89.64 * P$	0.34	57.11	133.44	0.050	0.01	0.58	0.00	Very poor
$CH_4VSadd = 556.04 - 6.69 * VS$	0.00	70.30	138.09	0.346	0.02	0.01	0.00	Very poor
$CH_4VSadd = 646.52 - 15.99 * K$	0.05	68.33	138.09	0,289	0.34	0.23	0.08	Very poor
Multivariate								
$CH_4VSadd = 327.81 + 8.46*HRT$	0.90	22.23	105.27	< 0.001	0.97	0.95	0.92	Great

 $R^2 = coefficient of multiple determination; RMSE = root mean square error; AIC = Akaike information criterion; p = F partial; index d = concordance index; r = correlation index; index c = confidence index; HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium; <math>CH_4 = \%$ methane; $CH_4 = \%$ methane; $CH_4 = \%$ methane production per kg of added VS; CODR = Chemical oxygen demand reduction.

Although significant, the contribution of N and P in the model was not enough to be included by the stepwise procedure in the multivariate model. Therefore, the best performance for predicting CH_4VS add was observed with the univariate linear regression using HRT (CH_4VS add = 327.81 + 8.46 * HRT), which presented the best values of R^2 , RMSE and AIC.

The univariate models for the reduction of chemical oxygen demand presented as significant variables (p = 0.05) the HRT and P. The values of R^2 , RMSE and AIC were respectively 0.90, 2.26 and 30.31 for HRT and 0.30, 6.01 and 61.4 for P (Table 4). As a result, the univariate model with the best performance for estimating the reduction of chemical oxygen demand (COD) was the one with HRT, since it had a better R^2 (0.90), when compared to P. Multivariate linear regression presented better results (CODR = 40.03 + 1.01 * HRT + 3.57 * P - 1.32 * K) for R^2 (0.91), RMSE (2.32) and AIC (35.9).

Jianbin Guo et al. (2013) studied the effect of temperature on the concentration of biomass digesters and treatment with pig manure, using semicontinuous digesters, and observed that the biogas production increased with the reduction of COD. The concentration of COD in the waste indicates the amount of organic matter in the effluent (Flotats et al. 2009).

Karmakar et al. (2010) reported that the reduction of organic material during digestion can be more affected by temperature and HRT than by the COD present on the digesting material. According to this, the conditions of the medium, when favorable to the degradation of organic matter, are more important for the COD removal efficiency than the attempt to obtain effluents with lower concentrations of COD.

Effluents with lower COD are also less concentrated and, consequently, decrease the biodigestion efficiency.

Table 5 presents data from the analysis of univariate and multivariate regression with methane production for the solid fraction separation (SFS) of the effluent. The CH_4 production showed a significant univariate model (p < 0.05) for HRT and N, respectively with R^2 , RMSE and AIC of 0.83, 0.56 and 14.45 and 0.29, 1.14 and 8.44. So, the HRT univariate model performed better in the estimation of methane content from SFS effluent. The best prediction ability was found for the multivariate linear regression ($CH_4 = 73.58 + 0.14 * HRT - 1.06 * N$), respectively with values of R^2 , RMSE and AIC of 0.88, 0.47 and 16.15.

Fdez-Güelfo et al. (2011) studied the influence of particle size of the organic matter using the organic fraction of municipal waste and synthetic waste. In both types of waste, they observed that univariate models with HRT presented the highest R^2 (0.99).

Analysis of univariate and multivariate regression of potential methane production per kg of VS added (CH₄ VSadd) presented only the HRT as a significant variable (p = 0.05). The significance of this variable in the models is related to the HRT correlation with other variables. Therefore, only the HRT (0.92) had high R^2 for predicting the production of methane contents (CH₄).

Fdez-Güelfo et al. (2011) studied two types of residues: one from organic fraction of municipal waste and other from synthetic waste. The two types of waste had univariate models in which the HRT showed the highest R² (0.99). Karmakar et al. (2010) pointed out that these conditions had more influence on the process than the specific characteristics of

Table 4. Univariate and multivariate models to estimate reductions of chemical oxygen demand from swine effluents with no solid fraction separation (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Model	R ²	RMSE	AIC	р	Index d	r	Index c	Performance
Univariate								
CODR = 51.77 + 0.87 * HRT	0.90	2.26	30.11	< 0.001	0.97	0.95	0.92	Great
CODR = 67.15 + 3.29 * TS	0.00	7.17	65.11	0.213	0.10	0.07	0.01	Very poor
CODR = 69.45 + 1.15 * N	0.00	7.18	65.11	0.349	0.05	0.04	0.00	Very poor
CODR = 113.14 - 8.60 * P	0.30	6.01	61.40	0.049	0.65	0.55	0.35	Very poor
CODR = 67.15 + 3.66 * VS	0.00	7.17	65.11	0.123	0.10	0.07	0.01	Very poor
CODR = 91.74 - 2.75 * K	0.15	7.22	64.36	0.055	0.52	0.40	0.21	Very poor
Multivariate								
CODR = 40.03 + 1.01 * HRT + 3.57 * P - 1.32 * K	0.91	2.32	35.90	< 0.001	0.98	0.95	0.93	Great

CODR = chemical oxygen demand reduction; R² = coefficient of multiple determination; RMSE = root mean square error; AIC = Akaike information criterion; p = F partial; index d = concordance index; r = correlation index; index c = confidence index; HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium.

substrates in digestion and that the HRT would be determinant for greater degradation of organic matter and also to easily stabilize the conditions of the medium for the production of methane.

The analysis of the TS reduction through the univariate models presented a significant effect (p < 0.05) for the variables HRT and TS (Table 6). The coefficient of determination (R^2), RMSE and AIC for

Table 5. Univariate and multivariate models to estimate methane content and potential of methane production from swine effluents with separation of solid fraction (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Model	R ²	RMSE	AIC	р	Index d	r	Index c	Performance
Univariate								
$CH_4 = 69.44 + 0.16 * HRT$	0.83	0.56	14.45	< 0.001	0.95	0.91	0.87	Great
$CH_4 = 73.39 + 0.07 * TS$	0.00	1.36	13.92	0.143	0.01	0.01	0.00	Very poor
$CH_4 = 83.43 - 2.78 * N$	0.29	1.14	8.44	0.040	0.64	0.54	0.34	Very poor
$CH_4 = 78.74 - 1.41 * P$	0.08	1.31	11.92	0.087	0.36	0.28	0.10	Very poor
$CH_4 = 75.03 - 2.79 * VS$	0.01	1.35	11.92	0.102	0.15	0.12	0.02	Very poor
$CH_4 = 73.00 + 0.06 * K$	0.00	1.36	13.86	0.213	0.08	0.06	0.00	Very poor
Multivariate								
$CH_4 = 73.58 + 0.14 * TRH - 1.06 * N$	0.88	0.47	16.15	< 0.001	0.97	0.94	0.91	Great
Univariate								
$CH_4 VSadd = 513.75 + 18.18 * HRT$	0.91	43.29	124.57	< 0.001	0.98	0.96	0.93	Great
CH_4 VSadd = 1028.07 - 36.43 * TS	0.00	148.65	164.05	0.112	0.05	0.03	0.00	Very poor
CH_4 VSadd = 1450.72 - 123.63 * N	0.08	112.96	154.55	0.078	0.00	0.41	0.00	Very poor
CH_4 VSadd = 1182.18 - 38.05 * P	0.02	68.94	137.84	0,096	0.00	0.15	0.00	Very poor
CH_4 VSadd = 1009.70 + 54.62 * VS	0.00	69.67	139.08	0,127	0.08	0.04	0.00	Very poor
$CH_4 VSadd = 986.42 + 6.94 * K$	0.02	69.10	137.84	0,0945	0.21	0.14	0.03	Very poor
Multivariate								
$CH_4 VSadd = 513.75 + 18.18 * HRT$	0.92	40.58	124.57	< 0.001	0.98	0.96	0.94	Great

R²= coefficient of multiple determination; RMSE = root mean square error; AIC = Akaike information criterion; p = F partial; index d = concordance index; r = correlation index; index c = confidence index; CH₄ = % methane; CH₄ VSadd = methane production per kg of added VS; HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium.

Table 6. Univariate and multivariate models to estimate reductions of total solid and volatile solid from swine effluents with solid fraction separation (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Model	R ²	RMSE	AIC	р	Index d	r	Index c	Performance
Univariate								
TSR = 46.60 + 0.56 * HRT	0.45	4.82	54.31	0.039	0.78	0.67	0.52	Fair
TSR = 24.89 + 25.82 * TS	0.34	5.28	57.28	0.045	0.70	0.58	0.41	Poor
TSR = 96.99 - 10.12 * N	0.17	5.92	60.90	0.056	0.52	0.41	0.21	Very poor
TSR = 87.99 - 7.28 * P	0.10	6.17	62.24	0.067	0.41	0.31	0.13	Very poor
TSR = 61.16 - 2.35 * VS	0.00	4.08	47.01	0.189	0.05	0.03	0.00	Very poor
TSR = 57.60 + 0.63 * K	0.04	3.99	47.01	0.098	0.29	0.21	0.06	Very poor
Multivariate								
TSR = 10.70 + 0.56 * HRT + 25.82 * TS	0.81	2.80	41.10	< 0.001	0.95	0.90	0.85	Very good
Univariate								
VSR = 51.38 + 0.62 * HRT	0.56	4.30	50.71	0.012	0.84	0.75	0.63	Medium
VSR = 36.88 + 21.90 * TS	0.24	5.68	59.57	0.051	0.62	0.49	0.30	Very poor
VSR = 104.28 - 10.33 * N	0.17	5.92	60.90	0.076	0.53	0.42	0.22	Very poor
VSR = 96.00 - 7.67 * P	0.11	6.16	62.18	0.089	0.44	0.33	0.14	Very poor
VSR = 67.70 - 2.37 * VS	0.00	4.06	46.89	0.234	0.05	0.03	0.00	Very poor
VSR = 64.66 + 0.56 * K	0.03	3.99	46.89	0.078	0.26	0.19	0.05	Very poor
Multivariate								
VSR = 20.94 + 0.62 * HRT + 21.90 * TS	0.83	2.71	39.84	< 0.001	0.95	0.90	0.86	Great

R²= coefficient of multiple determination; RMSE = root mean square error; AIC = Akaike information criterion; p = F partial; index d = concordance index; r = correlation index; index c = confidence index; TSR = total solids reduction; VSR = volatile solids reduction; HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium.

Table 7. Univariate and multivariate models to estimate reductions of chemical oxygen demand from swine effluents with solid fraction separation (Dourados, Mato Grosso do Sul State, Brazil, 2013).

Model	R ²	RMSE	E AIC	р	Index d	r	Index c	Performance
Univariate								
CODR = 54.52 + 0.40 * HRT	0.63	2.38	31.82	0.001	0.87	0.80	0.70	Good
CODR = 51.86 + 9.28 * TS	0.12	3.70	45.91	0.089	0.45	0.34	0.15	Very poor
CODR = 84.17 - 5.42 * N	0.13	3.68	45.66	0.077	0.45	0.36	0.16	Very poor
CODR = 91.53 - 7.34 * P	0.42	2.37	31.66	0.045	0.78	0.65	0.51	Fair
CODR = 73.83 - 17.60 * VS	0.11	2.95	38.51	0.084	0.45	0.33	0.15	Very poor
CODR = 61.31 + 0.36 * K	0.02	3.09	38.51	0.098	0.23	0.16	0.03	Very poor
Multivariate								
CODR = 56.43 + 0.38 * HRT + 7.85 * TS + 2.24 * N - 5.43 * P	0.89	1.33	20.29	< 0.001	0.97	0.94	0.91	Great

 R^2 = coefficient of multiple determination; RMSE = root mean square error; AIC = Akaike information criterion; p = F partial; index d = concordance index; r = correlation index; index c = confidence index; CODR = chemical oxygen demand reduction; HRT = hydraulic retention times; TS = total solids; N = nitrogen; P = phosphorus; VS = volatile solids; K = potassium.

HRT and TS were respectively 0.45 and 4.82, 54.41 and 0.34, and 5.28 and 57.28. The univariate model that showed the best performance in the estimation of methane content was the HRT, presenting a better R^2 (0.45), when compared to the TS (0.34). Multivariate linear regression (TSR = 10.70 + 0.56 * HRT + 25.82 * TS) presented R^2 , RMSE and AIC of 0.81, 2.80 and 41.1, respectively. There was no correlation between TS and HRT (Table 2).

The univariate models obtained for the reduction of SV presented significant effect (p < 0.05) only for the variables HRT and TS. The R^2 , RMSE and AIC for HRT and TS were respectively 0.56 and 4.30, 50.71 and 0.24, and 5.68 and 59.57 (Table 6). The univariate model that showed the best performance in estimating the reduction in VS was HRT, which had a higher R^2 , when compared to TS. For multivariate linear regression (VSR = 20.94 + 0.62 * HRT + 21.90 * TS), better results were obtained for R^2 (0.83), RMSE (2.71) and AIC (39.84).

The analysis of the reduction of COD through the univariate models presented a significant effect (p < 0.05) for HRT and P (Table 7). The coefficient of determination (R^2), RMSE and AIC for HRT and P were respectively 0.63 and 2.38, 31.82 and 0.42, and 2.37 and 31.66. In this way, the univariate model that performed better in estimating the reduction of chemical oxygen demand was the HRT, that had a higher R^2 , when compared to P. The multivariate linear regression (CODR = 56.43 + 0.38 * HRT + 7.85 * TS + 2.24 * N - 5.43 * P) presented R^2 , RMSE and AIC of 0.89, 1.33 and 20.29, respectively. There was significant correlation of P with ST and N, thus only the less correlated variables were included

in the model. The HRT was the only variable that contributed to the high R² for predicting changes of COD reduction.

CONCLUSION

The models in this study can be used for the accurate prediction of the main parameters of swine wastewater anaerobic digestion with and without separation of the solid fraction process. HRT is the factor that most affects the process of anaerobic digestion, being responsible for most of the variation captured by the presented models.

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