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by Keenan Keenan

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Evaluation and modelling of biogas production from batch anaerobic digestion of corn stover with oxalic acid

Abstract: Corn stover is one of the potential lignocellulosic biomasses as the raw material of biogas production. Pretreatment of lignocellulose substrates can enhance biodegradability and biogas yield. This study investigates the effect of oxalic acid pretreatment on biogas production during batch anaerobic digestion of corn stover. First-order, logistic, modified Gompertz and transference models predicted kinetic parameters during biogas production from pretreated corn stover. Results showed that oxalic acid pretreatment significantly affected biogas production ($p < 0.05$). The highest cumulative biogas yields of pretreated corn stover and untreated corn stover were 95.14 mL/gVS and 57.55 mL/gVS, respectively. Pretreated substrates improved biodegradability by 165%. Four kinetic models provided the determination coefficients R^2 higher than 0.9. The logistic model and modified Gompertz provided the best deviation of 1.57% and 3.75%, respectively. The logistic model proved the best fitting in predicting cumulative yields and simulating the kinetic model of anaerobic digestion of pretreated corn stover among the three models.

Keywords: kinetic model; first-order model; logistic model; modified Gompertz; transference model

INTRODUCTION

Anaerobic digestion (AD) is a biological process to produce biogas through organic material degradation by microbes without oxygen. Biogas composition consists of 50-70% CH_4 and 30-50% CO_2 with small components such as hydrogen sulphide, nitrogen, oxygen, siloxanes, volatile organic compounds (VOCs), carbon monoxide, and ammonia (Adnan et al., 2019). AD process can be divided into four stages, i.e., hydrolysis, acidogenesis, acetogenesis, and methanogenesis (Xu et al., 2019).

Corn stover belongs to lignocellulosic biomass since it is composed of cellulose (46.16 ± 0.46 wt%), hemicellulose (24.95±0.33 wt%) and lignin (5.90±0.21 wt%) (Yang et al.,

2022). Lignocellulose is a significant component of the plant cell wall composed of mainly cellulose, hemicellulose, and lignin (pram al., 2019). Anaerobic digestion can digest cellulose and hemicellulose, while lignin is a barrier against organic material degradation that inhibits the AD process (Abraham et al., 2020). Pretreatment is essential to enhance biodegradability, thus making the lignocellulosic biomass more accessible to microbes (Olugbemide et al., 2021). Lignocellulosic biomass has applied many pretreatment methods

Chemical pretreatment, especially acid pretreatment, reduces hemicellulose (Antonopoulou et al., 2020). Acid pretreatment can also condense and deposit lignin fraction; hydrolytic enzymatic activity runs well in acidic conditions (Dasgupta & Chandel, 2020). There are few studies on producing biogas from lignocellulosic biomass using chemical pretreatment. Amnuaycheewa et al. (2016) reported that pretreatment using 5.01% oxalic acid generates the highest biogas yield of 322.1 mL/g during the anaerobic digestion of rice straw. It is 7.40 times higher than untreated rice straw. Taherdanak et al. (2016) stated that pretreatment using dilute sulfuric acid for 120 min generates the maximum biogas yield of 513.9 mL/gVS during biogas production from wheat plants. The pretreated wheat plant obtains a 4% higher biogas yield than the untreated wheat plant. Jankovičová et al. (2022) revealed that pretreatment of 0.5% H₂SO₄ increased specific biogas production of rapeseed straw (by 71%) and wheat straw (by 32%). However, the kinetic model in producing biogas from corn stover using oxalic acid pretreatment has not been widely evaluated. Therefore, the study's objective was to evaluate kinetic models on batch anaerobic digestion of corn stover to obtain the best fitting biogas production curves and describe kinetic parameters using the First-Order, Logistic model, Modified Gompertz and Transference model. This study also investigated the effect of oxalic acid pretreatment on biogas production. A kinetic model of anaerobic digestion can be used to expect stability factors, types of reactors and substrates, and dynamic simulation of anaerobic digestion (Bakraoui et al., 2020).

MATERIAL AND METHODS

Feedstock preparation

Corn stover was collected from the fields in Yogyakarta. Corn stover was dried in the sun and ground into 1-2 mm using a grinder. Dried and ground corn stover was stored at room temperature before use. The fresh fluid rumen of the cow was obtained from a Slaughterhouse in Yogyakarta and used as inoculum.

Oxalic acid pretreatment

Chemical pretreatment was conducted using 10% (w/w) C₂H₂O₄ (oxalic acid) solution at room temperature for 6 hours. The pretreated corn stover was washed with distilled water and dried in the sun. The pretreated substrates were stored at room temperature until use.

Biogas production

The untreated and pretreated corn stover were mixed with inoculum and water to adjust a feed-to-inoculum ratio of 1. The substrates were loaded into a 1 L batch digester. Biogas production was carried out at room temperature for 30 days. Daily biogas volume was measured every three days using the water displacement method.

Kinetic analysis

The kinetic model predicts anaerobic digestion parameters such as the potential biogas production, biogas maximum rate and the lag phase time obtained from the experimental results (Khadka et al., 2022).

First-order kinetic model

This model assumes the hydrolysis step as a rate-limiting step in anaerobic digestion. The cumulative biogas yield is shown in Equation 1:

$$M = P_0 [1 - \exp(-kt)] \quad (1)$$

Where M is the cumulative biogas yield at time t (mL/gVS), P_0 is the methane potential of the substrate (mL/gVS), k is the first-order biogas production rate constant (1/day), t is digestion time (days)

Logistic model

The logistic model assumes the biogas production rate is proportional to the amount of biogas produced. This model fits an initial exponential increase and final stability at the highest production level. The logistic model is written in Equation (2)

$$M = \frac{P_0}{\left[1 + \exp\left\{\frac{4R_m(\lambda - t)}{P_0} + 2\right\}\right]} \quad (2)$$

Where R_m is the maximum methane production rate (mL/gVS/d), λ is the lag phase time (days).

Modified Gompertz model

This model illustrated the lag phase and the highest biogas production rate. The biogas production rate is supposed to be parallel to the specific growth of methanogens. The modified Gompertz model is given by Equation (3)

$$M = P_0 \times \exp\left\{-\exp\left[\frac{R_m \cdot e}{P_0}(\lambda - t) + 1\right]\right\} \quad (3)$$

Transference model

The transference model described the correlation between biogas production and microbial activity. It also analysed the anaerobic digestion process as the system's input and output signal of the system. It predicted the maximum biogas production based only on cumulative biogas over time. The model is presented in Equation 4

$$M = P_0 \times \left\{1 - \exp\left[\frac{-R_m(t - \lambda)}{P_0}\right]\right\} \quad (4)$$

Data analysis

The p-value was adjusted at 0.05, and the significance of the results was checked with p-values < 0.05, while no significant results were with p-values > 0.05 during the analysis of variance (ANOVA). The kinetic parameters were determined using non-linear regression by Solver in MS Excel.

The best-fit model can be identified through the highest R² coefficients and the smallest RMSE value. The deviation between experimental and predicted results can also be used to determine the best-fit model. The low deviation values (< 10%) suggest the accurate prediction of the model (Zahan et al., 2018).

Biodegradability

Biodegradability was determined by dividing cumulative biogas yields by theoretical biogas yields. The theoretical yield obtained from this study was 99.18 mL/gVS. It was estimated using the Buswell equation. The Equation to determine biodegradability is written below(Lahboubi et al., 2022):

$$\text{Biodegradability (\%)} = \frac{\text{cumulative biogas yield (mL/gVS)}}{\text{theoretical biogas yield (mL/gVS)}} \quad (5)$$

RESULTS AND DISCUSSION

Effect of oxalic acid pretreatment on biogas production

The effect of oxalic acid (C₂H₂O₄) pretreatment on biogas production was investigated using C₂H₂O₄ of 10%. Figure 1 presents the daily biogas yield for 30 days.

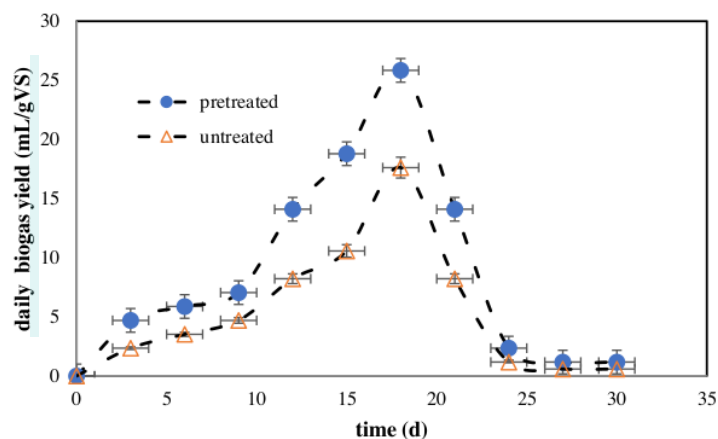


Fig 1. Daily biogas yield during anaerobic digestion of corn stover

Biogas production started on day 3 with biogas yields of 4.70 mL/gVS and 2.35 mL/gVS for pretreated and untreated substrates, respectively. Biogas yield then increased gradually until reaching peak yields of 25.84 mL/gVS and 17.62 mL/gVS on day 18 at the pretreated and untreated substrate, respectively. Biogas production then decreased regularly, with the lowest yields on day 30.

Results showed that adding $C_2H_2O_4$ positively affected biogas production, as shown in Figure 2.

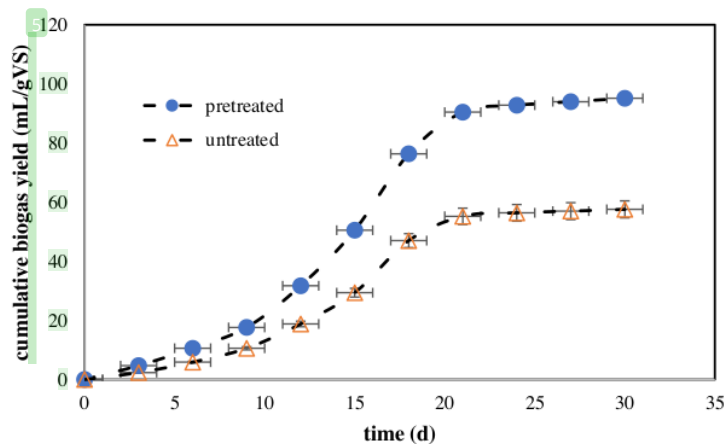


Fig. 2. Cumulative biogas yield during anaerobic digestion of corn stover

The pretreated substrate generated a higher cumulative yield of 95.14 mL/gVS than the untreated substrate (57.55 mL/gVS). Pretreatment using oxalic acid ($C_2H_2O_4$) could increase biogas yield by 65%. This result occurs because oxalic acid can hydrolyse hemicelluloses during lignocellulose pretreatment (Cheng et al., 2018). Deng et al. (2016) also stated that oxalic acid had high selectivity for hemicellulose degradation. The break of hemicellulose content can increase the degradability of substrate; as a result, biogas production also increases (Phutela & Sahni, 2012). The previous result also reported that biogas production increased by 61.87% during the anaerobic digestion of water hyacinth using oxalic acid pretreatment (Tantayotai et al., 2019). Statistical analysis also verified that oxalic acid pretreatment had a significant effect on biogas production with a p-value of 0.0149 ($p < 0.05$)

Biogas production kinetic using a first-order model

The first-order model fitted the cumulative biogas yields of anaerobic digestion from pretreated corn stover. The comparison between experimental results and the model is presented in Figure 3.

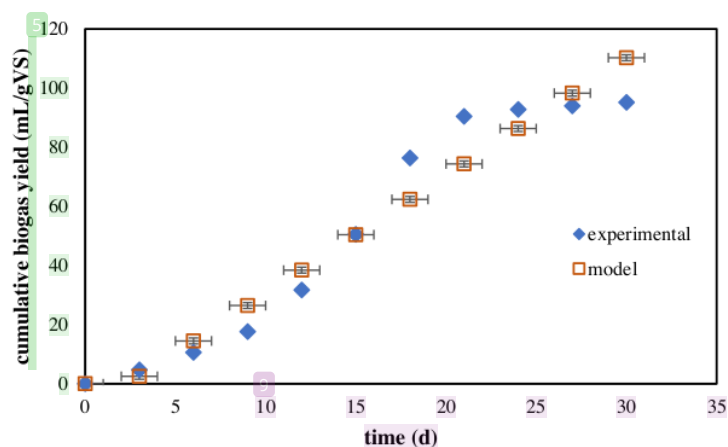


Fig. 3. Biogas kinetic model using the first-order model during anaerobic digestion of pretreated corn stover

The kinetic constant (k) was found to be $0,1123 \text{ day}^{-1}$. The simulation results were 0.9302, 0.9125, 8.9793, and 9.2371×10^{-14} for R^2 , adjusted R^2 , RMSE, and SSE values, respectively. The methane potential of the substrate obtained from the first-order model was 6.103 mL/gVS. The cumulative yields' deviation from experimental results and model was $\pm 13.71\%$. The result showed that the first-order model gave a good fit in expressing cumulative biogas yield because the R^2 value obtained from the first-order model was higher than 0.9. Previous results conducted by Pečar & Goršek (2020) and Nweke et al. (2022) also stated that the first-order model was an excellent fit to predict the kinetic of anaerobic digestion with $R^2 > 0.9$.

Biogas production kinetic using the logistic model

Figure 4 compares cumulative yields obtained from the logistic model and experimental results. The simulation model obtained R^2 , adjusted R^2 , SSE, and RMS values of 0.9452, 0.9384, 3.1964, and 2.3029, respectively.

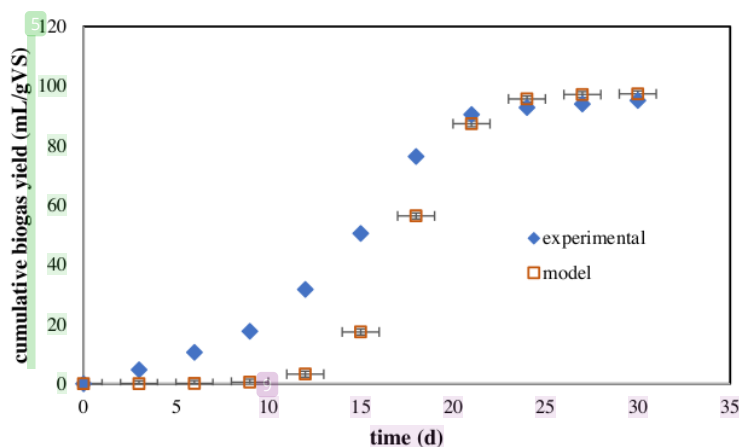


Fig. 4. Biogas kinetic model using the logistic model during anaerobic digestion of pretreated corn stover

The cumulative yield difference between experimental results and the logistic model was $\pm 1.57\%$. The R^2 value obtained from the logistic model was higher than 0.9, which

indicated that the model could become a suitable model for predicting biogas kinetic from pretreated corn stover. Compared to the R^2 value obtained from the first-logistic model, the logistic model had a higher R^2 value. The logistic model had a better simulation than the first-order model. The lag phase time (λ) of relevant results was 14.22 days. The maximum methane production rate (R_m) and the methane potential of the substrate (P_0) obtained from the logistic model were 14.9614 mL/gVS/day and 97.419 mL/gVS, respectively. A previous result also reported that the logistic model gave the R^2 value higher than 0.9 on kinetic modelling of biogas production from poultry slaughterhouse wastes (Ware & Power, 2017). Gong et al. (2019) also found the R^2 coefficients > 0.9 during a kinetic analysis of the anaerobic digestion of sewage sludge using a logistic model.

Biogas production kinetic using a modified Gompertz model

The modified Gompertz equation was used to model cumulative biogas yield, as presented in Figure 5. The cumulative biogas yield obtained from the experimental results and the modified Gompertz model had a difference of $\pm 3.75\%$.

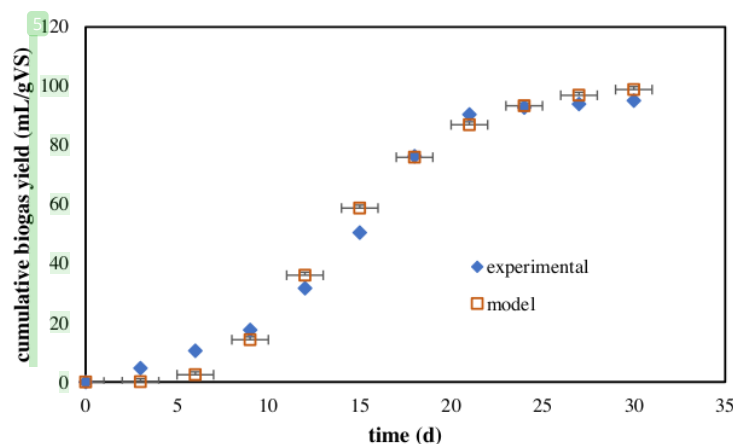


Fig. 5. Biogas kinetic model using modified Gompertz model during anaerobic digestion of pretreated corn stover

The simulation results gave the R^2 , adjusted R^2 , SSE, and RMSE values of 0.9416, 0.9344, 0.0013, and 4.4800, respectively. The R^2 value obtained from the modified Gompertz

was higher than 0.9, signifying the excellent fit of the modified Gompertz in calculating the accumulation process of biogas yields. The prior study (Zahan et al., 2018) also found that the modified Gompertz gave the $R^2 > 0.9$ in kinetic modelling of the anaerobic digestion of agricultural wastes.

The R^2 value fitted by the modified Gompertz was higher than R^2 fitted by the first-order model, indicating the modified Gompertz model predicted cumulative yields more fitted than the first-order model. However, R^2 obtained from the modified Gompertz was lower than R^2 fitted by the logistic model, denoting that the modified Gompertz was less accurate to be applied to the kinetic model of biogas production from pretreated corn stover. The kinetic parameters resulted in the lag phase time (λ) of 12.13 days, maximum methane production rate (R_m) of 2.9245 mL/gVS/day, and the methane potential of the substrate (P_0) of 37.1681 mL/gVS.

Biogas production kinetic using the transference model

Figure 6 shows cumulative biogas yields between the experimental results and the transference model. The simulation of the transference model provided a high R^2 value of 1, followed by an adjusted R^2 of 1, an SSE value of 6.4551, and an RMSE value of 2.7369. The experimental and model cumulative biogas yields had a difference of $\pm 12.69\%$.

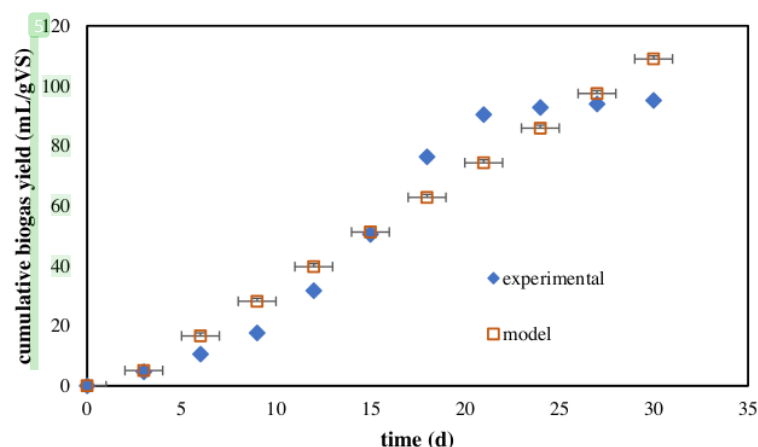


Fig. 6. Biogas kinetic model using transference model during anaerobic digestion of pretreated corn stover

The transference model had the highest R^2 value among the three models (the first-order model, the logistic model, and the modified Gompertz model). The kinetic parameters observed by the transference model were the λ value of 1.63 days, the R_m value of 3.8476 mL/gVS/day, and the P_0 value of 2.5547×10^8 mL/gVS. The R^2 coefficient of 1 signifies that the regression model expresses all predicted variables, which means that the relationship between measured and predicted variables is perfect (Jierula et al., 2021). The prior study reported by Ali et al. (2018) also found that the logistic model obtained R^2 values > 0.9 during the kinetic analysis of the anaerobic digestion of cow manure.

Table 1 summarises the kinetic parameters obtained from the first-order, modified Gompertz, and transference models.

As seen in Table 1, all models proposed for the kinetic simulation were a good fit for predicting the cumulative yields due to the R^2 values > 0.9 . Though the transference model had the perfect R^2 of 1 and a relatively small RMSE value, the deviation of this model was higher than 10%; thus, the kinetic model simulation does not recommend the transference model as the best-fit model. The best-fit model suggested for kinetic modelling of anaerobic

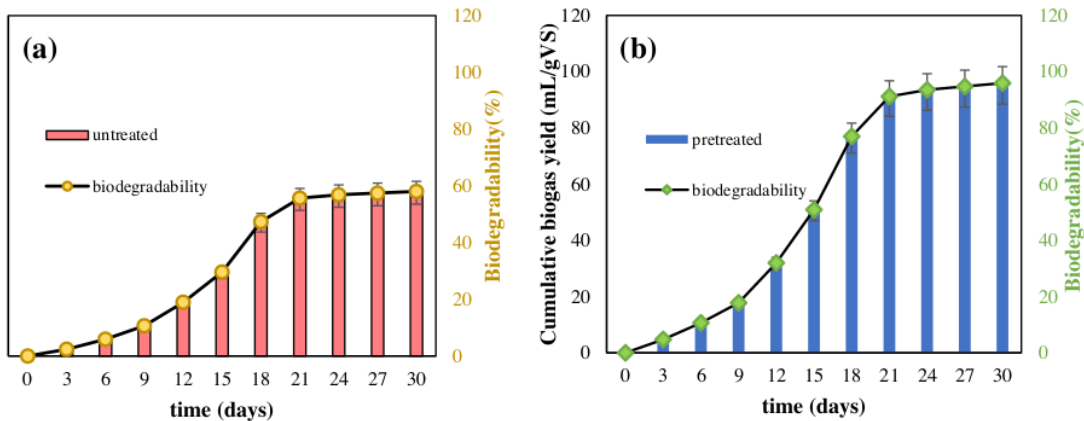
digestion of pretreated corn stover was the logistic model because it had the smallest RMSE and the lowest deviation among the other models. The R^2 value obtained from the logistic model also met the accuracy of the regression model due to the $R^2 > 0.9$

Table 1. Summary of kinetic parameters

Parameters	Units	first-order model	logistic model	modified Gompertz model	transference model
Methane potential of the substrate (P_0)	mL/gVS	6.1030	97.419	37.1681	2.54×10^8
Maximum methane production rate (R_m)	mL/gVS/d	Not calculated	14.9614	2.9245	3.8476
Lag phase time (λ)	d	Not calculated	14.23	12.13	1.68
The rate constant (k)	1/d	0.1123	Not calculated	Not calculated	Not calculated
R^2		0.9302	0.9452	0.9416	1
Adjusted R^2		0.9215	0.9384	0.9344	1
SSE		9.23×10^{-14}	3.9164	0.0013	6.4551
RMSE		8.9793	2.3029	4.4800	2.7369
Difference between measured and predicted biogas yield	%	13.71	1.57	3.75	12.69

Biodegradability

Figure 7 presents biodegradability on untreated and pretreated substrates.



1 Fig. 7. Cumulative biogas yield and biodegradability as the function of (a) untreated substrate; (b) pretreated substrate

Pretreated and untreated substrates provided the highest biodegradability of 95.93% and 58.03%, respectively. Pretreated substrate increased biodegradability by ± 165%. As seen in Figure 7, the higher cumulative yields generated higher biodegradability. Pretreated substrates had a higher biodegradability than untreated substrates. This result 8 indicated that pretreatment could improve biogas production from corn stover.

CONCLUSION

Oxalic acid pretreatment significantly affected biogas yields with a p-value < 0.05. Pretreated substrates had higher biodegradability (95.93%) than untreated substrates (58.03%).

Cumulative yields obtained from pretreated corn stover increased by 65%. The result of the kinetic analysis showed that the determination coefficients R^2 obtained from all models were higher than 0.9. All four models could describe the kinetic of anaerobic digestion from pretreated corn stover. 2 According to the RMSE values and the difference between the experimental and predicted values, it was suggested that the logistic model was more accurate and a better fit than the first-order model, modified Gompertz model, and transference model 3 in fitting experimental biogas yields.

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