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# Assessing the Sensitivity of Technical Performance of three Ethanol Production Processes based on the Fermentation of Steel Manufacturing Offgas, Syngas and a 3:1 Mixture Between H<sub>2</sub> and CO<sub>2</sub>

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## Abstract

This study assesses the sensitivity of the technical, environmental and economic performance of three ethanol production process based on the fermentation of three gas mixtures: i) CO-rich flue gas from steel manufacturing, ii) biomass-based syngas with a H<sub>2</sub>/CO ratio of 2 and iii) a 3:1 combination between H<sub>2</sub> and CO<sub>2</sub>. The sensitivity analysis is based on stochastic bioreactor simulations constructed by randomly generated combinations of eight parameters that command the fermentation process i.e., temperature, pressure, gas feed dilution with an inert components, ethanol concentration, height of the liquid column, mass transfer coefficients, superficial gas velocity and, acetic acid co-production.

The sensitivity analysis identified that the bioreactor technical performance is highly sensitive to variations on pressure, liquid column height and the mass transfer coefficients. The pressure mainly improves mass transfer and consequently ethanol productivity whereas liquid column height improves the gas residence time and consequently the efficiency in the gas utilization. The trend was common for the three gas supply options. The results suggested that in order to produce an optimal bioreactor design, there are options to optimize the productivity and the gas utilization simultaneously.

The results from the sensitivity analysis may help guiding a subsequent multi-objective process optimization study.

**Keywords:** ethanol, syngas fermentation, sensitivity analysis, stochastic simulation.

## 1. Introduction

Prevention of global warming is currently pushing global policy-making towards the reduction of CO<sub>2</sub> emissions, which are mostly derived from the combustion of fossil fuels (Boden et al., 2013). Lignocellulosic biomass is seen as an alternative sources of fuels and chemicals that during their life cycle may result in lower carbon emissions (Liu et al., 2017). These feedstocks are abundant and renewable and can be thermochemically converted into gas mixtures containing mainly CO, H<sub>2</sub>, CO<sub>2</sub> (Heidenreich and Foscolo,

2015; Matsakas et al., 2017). The gas is commonly referred as syngas and can be used as feedstock for fermentations (Kundiya et al., 2010).

Only limited details about the industrial process performance have been made public. What has been reported is that microbial selectivity for ethanol falls around 95 % and that gas utilization overpasses 90 % (Simpson, 2018); additionally, it is argued that the exist an energy surplus generated from the based fermentation of syngas (Handler et al., 2016). Moreover, the claims that ethanol concentration in the fermentation media could be held above 50 g/L and overhead pressure should not overpass 3 atm (Li et al., 2017; Trevethick et al., n.d.) have been patented.

In consequence, a mathematical model was developed and reported elsewhere [Almeida, forthcoming] to simulate ethanol production in a bubble column bioreactor fed by CO, H<sub>2</sub> and CO<sub>2</sub> mixtures. That model is used to quantify the sensitivity of the bioreactor technical performance to certain parameters that command the fermentation step. The assessment is applied to three different gas feed compositions.

## 2. Methodology

### 2.1. Process configurations

The three process configurations which have been considered in this study differ on the gas production processes: CO-rich BOF offgas from the steel-manufacturing process; *ii*) a 3:1 mixture between H<sub>2</sub> and CO<sub>2</sub>, and *iii*) syngas with a H<sub>2</sub>/CO ratio of 2.

The fermentation process consists on a bubble column bioreactor fed by the gas mixture. The ethanol produced inside the fermentor is at all times given two possible exit routes *i.e.* *i*) pre-concentrated along the offgas, where it is subsequently condensed and recovered from by flash separation, and *ii*) along a liquid broth outflow. Acetic acid, also exits the bioreactor along the liquid outflow. The alcohol is distilled out of the two streams by atmospheric distillation and finally dehydrated. The unconsumed gas is here treated as waste and combusted before being released into the atmosphere, as proposed by Handler et al, 2016 (Handler et al., 2016).

### 2.2. Simulation of the fermentation processes

The simulation of the bioreactor uses a model previously presented elsewhere [Almeida, forthcoming]; therefore, only the basic structure of such model is introduced here.

#### 2.2.1. Simulation of the fermentations of BOF offgas and the H<sub>2</sub>/CO<sub>2</sub> mixture

The fermentation model for these two cases is formed by *i*) a black-box model of the main reactions carried out by acetogenic bacteria and *ii*) a mass transfer model of the large bubble column bioreactor. The stoichiometry of the microbial metabolic reaction is constructed by the combination of the catabolic and anabolic reactions.

Ethanol and acetic acid are the products of catabolism, thermodynamically powered by the uptake of the electron donors, CO and H<sub>2</sub>. Cells are produced during anabolism. The Gibbs free energy harvested from the electron donors (CO and H<sub>2</sub>) during catabolism powers cells. The uptake of CO and H<sub>2</sub> are assumed to follow hyperbolic kinetics. The maximum uptake rate of the electron donors is estimated using thermodynamics (Heijnen, 2013).

Mass transfer in the bioreactor is simulated assuming the process operates in continuous mode at steady-state (Heijnen and van't Riet, 1984; van't Riet and Tramper, 1991). Mass transfer is driven by the energy input provided by the gas sparging [42]. A system of non-linear equations formed by the mass balances is solved by a constrained optimization of the volumetric ethanol productivity ( $R_{et}$ ).

*2.2.2. Simulation of the fermentation of biomass-derived syngas*

Since the black-box model of microbial reactions is only able to simulate the consumption of either CO or H<sub>2</sub>, the simulation of biomass-derived syngas (BDS) consumption is done indirectly adding the mass and energy streams contributions from CO fermentation (the BOF offgas case) and H<sub>2</sub>/CO<sub>2</sub> fermentation cases. The H<sub>2</sub>/CO ratio in syngas is assumed to be 2.

*2.3. Process assessment and performance indicators*

The ethanol production processes are evaluated from the perspective of the bioreactor technical performance through the ethanol volumetric productivity ( $R_{et}$ ) and gas utilization ( $U_S$ ), defined as the percent change on the gas molar flow rate of electron donors across the fermentor.

*2.4. Stochastic simulation of the bioreactor*

The operation of the bioreactor is simulated under 5000 randomly generated combinations of eight input parameters, which are considered to command the bioreactor performance *i.e.*, *i*) process temperature ( $T$ ), *ii*) top reactor pressure ( $p_t$ ), *iii*) gas feed dilution ( $f_{Dil}$ ), *iv*) maximum ethanol concentration ( $C_{et}^{max}$ ), *v*) liquid column height ( $h_L$ ), *vi*) mass transfer coefficient factor ( $f_{kLa}$ ), *vii*) acetic acid production factor ( $f_{Ac-}$ ) and *viii*) the pressure-corrected superficial gas velocity ( $v_{sG}^c$ ). See Table 1.

**Table 1** Maximum and minimum values used the input parameters in the stochastic bioreactor simulation

<b>Input parameter</b>	$T$ [°C]	$p_t$ [atm]	$f_{Dil}$ [%vol.]	$C_{et}^{max}$ [g/L]	$h_L$ [m]	$f_{kLa}$ [-]	$f_{Ac-}$ [%]	$v_{sG}^c$ [m/s]
Minimum value	27	0.5	0	30	8	0.5	0	0.07
Maximum value	67	3.5	45	120	64	2.0	0.15	0.14

*2.5. Sensitivity analysis*

The sensitivity of bioreactor and overall process performance indicators (or output parameters - OP) is evaluated using standardized regression coefficients since it offers “a good approximation to a global sensitivity with affordable computational demand” (Morales-Rodriguez et al., 2012), and allows establishing a hierarchical classification of the model IP’s according to the level of impact on a determined OP. This method implies that process performance has a linear relation with each input parameter (IP). The reliability of the regression coefficients is evaluated using coefficients of determination ( $R^2$ ).

**3. Results and discussion**

*3.1. Distribution trends in the performance indicators*

The distribution of the bioreactor performance indicators corroborates previous observations suggesting that the fermentation of H<sub>2</sub>-rich gases may lead to higher productivity and gas utilization than the fermentation of CO-rich gases (Noorman and Heijnen, 2017) (Almeida et al., forthcoming) (see Figure 1). The observation is sustained on the fact that H<sub>2</sub> mass transfer to the liquid is faster than CO transfer; since mass transfer is linearly related for bioreactor productivity, then a higher mass transfer rate means higher productivity. Similarly, a higher productivity means that the gas consumption inside the bioreactor is more efficient, and therefore, gas utilization is higher.

Since BDS fermentation is simulated using the respective contributions of CO and H<sub>2</sub>/CO<sub>2</sub> fermentations, then it is reasonable that bioreactor performance lies between the

performances obtained for BOF offgas and H<sub>2</sub>/CO<sub>2</sub> fermentations. In addition, as the H<sub>2</sub>/CO ratio in the syngas is 2, then the performance of the syngas fermentor falls closer to that of the H<sub>2</sub>/CO<sub>2</sub> fermentor than to the CO fermentor.

By comparing bioreactor performances with other study (Almeida et al., forthcoming) there is a high probability that gas utilization may be improved far more than bioreactor productivity. This observation is based on the fact that while the median gas utilization

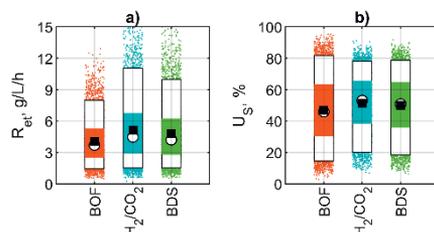


Figure 1 Boxplots summarizing the distribution of a) ethanol productivity and b) gas utilization.

On the boxplots: the colored vertical rectangles represent the extension of the 25th and 75th percentiles; the white vertical rectangles represent the extension of the 5th and 95th percentiles; the small colored dots represent the outliers; the white circles represent the median values; and the black square boxes represent the mean values.

values are 45 and 53 % for CO and H<sub>2</sub>/CO<sub>2</sub> fermentations, respectively, the median ethanol productivities are 3.7 and 4.5 g/h/L for CO and H<sub>2</sub>/CO<sub>2</sub> fermentations, respectively. However, this observation does not necessarily suggest that productivity cannot be further increased, as there is a 25 % probability (see 25<sup>th</sup> and 75<sup>th</sup> percentiles in Figure 1) that  $R_{et}$  may increase to 5.3, 6.8 and 6.2 g/L/h for BOF offgas, H<sub>2</sub>/CO<sub>2</sub> and BDS fermentations, respectively. In addition, there is a 5 % probability (see 5<sup>th</sup> and 95<sup>th</sup> percentiles in Figure 1) that  $R_{et}$  may be further improved to 8, 11 and 10 g/L/h for the same three gas supply options. That would be an improvement between 2 and 2.5 times from a previous report (Almeida et al., forthcoming).

Similarly, gas utilization in the bioreactor could be as high as 63, 66 and 65 % for BOF offgas, H<sub>2</sub>/CO<sub>2</sub> and BDS fermentations, respectively with a 25 % probability. Gas utilization could climb to 82, 78 and 79 % for the same gas supply options with a 5 % probability. It is therefore, more probable that the BOF offgas fermentation case has a better gas utilization than the fermentation of H<sub>2</sub>-rich gases.

In addition, although it may be encouraging to see that gas utilization could surpass 90 % as previously reported by LanzaTech, such achievement is highly unlikely with the proposed bioreactor configuration as there is only a 5 % probability that  $U_S$  may increase beyond 80 % (see Figure 1).

### 3.2. Sensitivity analysis by the standardized regression coefficients

Table 2 shows the standardized regression coefficients ( $\beta_{IP_i}$ ) of ethanol volumetric productivity and gas utilization in the bioreactor. The value of  $R^2$  is also included in Table 2 to show which OP's had an acceptable linear relation with all the IP's.

The relation between  $R_{et}$  and  $U_S$  and all the IP's is acceptably linear for the three gas supply cases. Remarkably, the IP's hierarchical classification is also common for these two OP's in three gas supply cases. Considering that the sign of the  $\beta_{IP_i}$  indicates whether the influence of each IP on one OP is positive or negative,  $R_{et}$  may be largely improved by  $p_t$  and the  $f_{kLa}$  and in a lower level of influence by the  $v_{SG}^c$ . On the other hand, the most detrimental IP for  $R_{et}$  is the gas feed dilution. According to [Almeida, forthcoming],

these four IP's are deeply related to the rate of mass transfer in the bioreactor.  $f_{kLa}$  and  $v_{sG}^c$  determine the value of the mass transfer coefficients while  $p_t$  and  $f_{Dil}$  determine the value of the partial pressures of the electron donors in the gas phase and therefore their saturation concentrations in the liquid phase and thus, their driving forces for mass transfer. While increases in  $p_t$  will widen the mass transfer driving force, increases in dilution will shrink the driving force.

Table 2 Standardized regression coefficients ( $\beta_{IP_i}$ ) relating the model outputs with each input parameter

OP	Gas supply case	IP								R <sup>2</sup>
		T	p <sub>t</sub>	f <sub>Dil</sub>	C <sub>et</sub> <sup>max</sup>	h <sub>L</sub>	f <sub>kLa</sub>	f <sub>Ac</sub> <sup>-</sup>	v <sub>sG</sub> <sup>c</sup>	
R <sub>et</sub>	BOF	0.13	0.57	-0.36	0.02	0.14	0.54	-0.06	0.32	0.90
	H <sub>2</sub> /CO <sub>2</sub>	0.07	0.61	-0.43	0.00	0.07	0.45	-0.05	0.28	0.86
	BDS	0.09	0.60	-0.41	0.00	0.09	0.48	-0.05	0.29	0.88
U <sub>S</sub>	BOF	0.18	-0.05	0.01	0.01	0.77	0.55	0.00	-0.13	0.95
	H <sub>2</sub> /CO <sub>2</sub>	0.11	0.15	-0.04	0.01	0.77	0.54	0.00	-0.13	0.92
	BDS	0.14	0.08	-0.02	0.01	0.78	0.55	0.00	-0.13	0.95

Gas utilization, in turn may be mostly improved by  $h_L$ ,  $f_{kLa}$  and  $T$ , while  $v_{sG}^c$  will be slightly detrimental. The sensitivity of  $U_S$  to  $h_L$  is remarkably high as the  $\beta_{IP_i}$  is the closest to one, the maximum possible value. This relation is due to the fact that as the liquid column height increases, so does the residence time of the gas inside the bioreactor. On the other hand,  $v_{sG}^c$  offers the opposite effect: as the gas velocity through the bioreactor increases, the residence time decreases. In addition, the influence of  $f_{kLa}$  and  $T$  over  $U_S$  could be regarded as a side effect of their positive influence over  $R_{et}$ , which means that the mass transfer increases, the gas consumption is more efficient.

The influence of ethanol concentration over  $R_{et}$  and  $U_S$  is negligible according to the estimated  $\beta_{IP_i}$ ; the lack of mathematical connection between this IP and the two specified OP's (Almeida et al., forthcoming) might be the cause of this negligible influence of  $C_{et}^{max}$  over bioreactor performance. Thus, regarding the fact that possible inhibition by ethanol is not considered by the electron uptake kinetic expressions in the black-box model of microbial reactions, the negligible influence of  $C_{et}^{max}$  might be somewhat underestimated. The similarity between the sensitivities for the three gas supply cases is caused by the fact that the operation of the bioreactor does not differ significantly when either CO or H<sub>2</sub> are the electron donors (Almeida et al., forthcoming). However, the intensity of such sensitivities differ between the three cases. For example, BOF offgas fermentation  $R_{et}$  is the most benefited indicator when the mass transfer coefficient factor increases. Similarly, due to the inhibition of CO at high partial pressures, the bioreactor pressure has a negative effect over gas consumption in the BOF offgas case while pressure is mostly beneficial for the fermentation of H<sub>2</sub>-rich gases.

#### 4. Conclusions

The present study showed that ethanol volumetric productivity could be as high as 8, 11 and 10 g/L/h for the fermentation of BOF offgas, H<sub>2</sub>/CO<sub>2</sub> and biomass-derived syngas, respectively. This increase may be achieved by a combination of high mass transfer coefficients, high top bioreactor pressure and large gas flow rates across the bioreactor.

These three IP's improve mass transfer rates and therefore ethanol productivity. In addition, the dilution of the gas feed is to be avoided since it has the most negative effect over productivity. The IP's that increase the productivity also have a positive impact on gas utilization. However, gas utilization may be mainly improved by tall bioreactors where the residence times are higher.

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