Dynamic modeling and controllability analysis of an ethanol reformer for fuel cell application

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Abstract

This work presents a controllability analysis of a low temperature ethanol reformer based on a cobalt catalyst for fuel cell application. The study is based on a nonlinear dynamic model of a reformer which operates in three separate stages: ethanol dehydrogenation to acetaldehyde and hydrogen, acetaldehyde steam reforming, and water gas shift reaction. The controllability analysis is focused on the rapid dynamics due to mass balances and is based on a linearization of the complex non-linear model of the reformer. RGA, CN and MRI analysis tools are applied to the linear model suggesting that a good performance can be obtained with decentralized control for frequencies up to 0.1 rad/s.

1. Introduction

Ethanol is a promising source of hydrogen as it is a renewable raw material when is obtained from biomass, and hence, research on catalytic steam reforming of ethanol and ethanol reformers is acquiring increasing interest. Ethanol reforming, as illustrated in Eq. (1), has been extensively studied over catalysts based on Ni, Ni/Cu, Co, and noble metals (Pd, Pt, Rh and Ru) [1, 2]. The reaction is reversible and highly endothermic ($\Delta H_{298} = 208.4$ kJ mol$^{-1}$), which accounts for the requirement of operation temperatures usually above 873K. At such high temperatures ethanol is mainly reformed into a mixture of H$_2$ and CO, as shown in Eq. (2). As carbon monoxide is a strong poison for the anode’s platinum catalyst of the fuel cell it is mandatory to pass the reformate through a water gas shift reactor [3-4], see Eq. (3), in order to generate further hydrogen.
and eliminate CO. However, cobalt-based catalysts are particularly suitable for ethanol steam reforming at lower temperatures, 623–673K, where the water gas shift reaction is very effective. For this reason, increasing attention is being focused on developing low-temperature catalytic processes with cobalt catalysts, both in packed reactors [5-6] as well as in microreactors for portable applications [7-8].

\[
C_2H_5OH + 3 H_2O \rightarrow 6 H_2 + 2 CO_2 \tag{1}
\]

\[
C_2H_5OH + H_2O \rightarrow 4 H_2 + 2 CO \tag{2}
\]

\[
CO + H_2O \leftrightarrow CO_2 + H_2 \tag{3}
\]

In a previous contribution, the authors reported results addressing the dynamic modeling of a three-module reactor for fuel cell hydrogen feeding [9]. Three specific catalysts were selected for each of the three modules of the reforming unit. Nanocrystalline SnO2 was used for the first step comprising ethanol dehydrogenation, while acetaldehyde steam reforming was performed in the second module with a Co(Fe)/ZnO catalyst doped with Na\(^+\). For the final water-gas shift step a reported kinetics for a commercial catalyst based on Fe\(_2\)O\(_3\)-Cr\(_2\)O\(_3\) was selected [11]. Detailed kinetic experiments over well-defined samples for the first two stages of the process were reported, as well as fitted parameters for power-law type kinetic expressions to quantify the correspondent reaction rates. A dynamic mathematical model of the three-stage reformer was presented as a tool for design of control-oriented devices. Monoliths were chosen to support the referenced catalysts. Fig. 1 shows a schematic representation of the three-staged reforming process.
The present contribution exploits the mathematical model presented in the above-cited reference [9] to perform a controllability analysis of the reforming unit. While the overall aim of the project is the design of a control strategy of the complete system (reformer + fuel cell), this paper focuses on the analysis of the fast response of the reformer due to changes in flow/concentrations of the feed. Future work will address the design of temperature controllers and the integration of the reformer with the fuel-cell. The following sections comprise a description of the mathematical model of the reformer and the analysis methodology followed by sensitivity and controllability analyses.

2. Non linear Model

A one-dimensional, pseudo-homogeneous, non-steady-state model has been used to represent the dynamic behavior of the ethanol reforming reactor in the already referenced series of monolithic stages. 1-D pseudo-homogeneous models are usually selected for control-oriented applications to reduce the solving time of the equations system, [10]. The kinetic expressions for the reactions occurring in stages 1 and 3 are taken from [9] and [11], respectively. However, a modification has been introduced in the present work regarding the system of reactions over the cobalt-based catalyst in bed 2. In fact, a parallel-series scheme is proposed to adjust the experimentally-observed product distribution, where acetaldehyde is reformed with steam towards carbon monoxide or carbon dioxide, which are consecutively balanced by the water-gas shift reaction:

\[ \text{C}_2\text{H}_4\text{O} + \text{H}_2\text{O} \rightarrow 3 \text{H}_2 + 2 \text{CO} \]  

(4)
First-order, power-law type kinetics were used for the three reactions. For reactions (4) and (5) a dependency on only acetaldehyde concentration is considered; kinetics for the water-gas shift reaction, Eq. (6), is based on the partial pressures of all the components involved in the reaction. By using a non-linear multi-parametric fitting routine, we obtained values of the reaction constants (at 673K) of $1.38 \times 10^4 \text{ mL g}_{\text{cat}}^{-1} \text{ h}^{-1}$, $1.52 \times 10^4 \text{ mL g}_{\text{cat}}^{-1} \text{ h}^{-1}$ and $9.70 \times 10^2 \text{ mol m}^{-3} \text{ s}^{-1} \text{ atm}^{-2}$, respectively.

The mass balance used to represent the non-steady-state reactor behavior, along with the corresponding initial and boundary conditions, is presented below:

Mass balance (component $j$)

$$\frac{\partial C_j}{\partial t} + \frac{\partial (vC_j)}{\partial z} = v_j r_i$$  \hspace{1cm} (7)

Initial conditions          $C_j(0,x) = C_{j,0}(x) \quad \forall x \in [e,L], \ j = 1...N$  \hspace{1cm} (8)

Boundary conditions        $C_j(t,0) = C_{j,e}(t) \quad \forall t > 0$  \hspace{1cm} (9)

The major assumptions underlying the model are the following:

- Gas properties are function of temperature and gas concentration. Ideal gas law is applicable due to the low operating pressure of ca. 1 bar.
• The pressure drop in the reactors is assumed negligible due to the high void fraction of the monolithic structures.

• The use of low-diameter monoliths and optimized inlet distributors supports the assumption of 1-D model avoiding the occurrence of pronounced radial mass profiles.

The numerical solution of the set of partial differential equations, Eq. (7) to (9), is accomplished by its transformation into an ODE-system by discretization of the spatial derivative. To this end, backward finite differences have been selected (first-order, 38 equidistant discretization points). The resultant ODE equations (198th order) are solved by an algorithm implemented in MATLAB® (ODE45 Normand-Prince). Additional details regarding the mathematical model and its solving strategy can be found in [9].

3. Methodology

We consider the complete nonlinear model of the three-staged reforming process with two inputs and 6 outputs:

• Inputs: \(F_{C_2H_5OH,e}, F_{H_2O,e}\)

• Outputs: \(F_{C_2H_5OH,L}, F_{C_2H_4O,L}, F_{H_2O,L}, F_{H_2,L}, F_{CO,L} \text{ and } F_{CO_2,L}\)

In this work, firstly a steady state sensitivity analysis is performed on the complete nonlinear model around a nominal operating point \(OP_1\). This operating point has been selected from appropriate conditions of the catalysts (which correspond to \(T_{s1} = 648K\), \(T_{s2} = 673K\) and \(T_{s3} = 633K\)) and the dimension of each monolith to provide the required
hydrogen flow to feed a 1kW PEM fuel cell. See [9] for more details. The static input-output non linearity characteristic is verified for different input levels.

One of the main goals of this work is to perform the controllability analysis of the reformer system. To this end, an accurate study of linearized versions of the model, at different operating points, is considered. The resulting MIMO linear systems are studied using different analysis tools. These tools are mathematic operators applied to the squared transfer functions of the linear system that provide relevant information such as stability, controllability, sensitivity, robustness, etc. They are applied to the process, without control, in order to characterize the controllability of the system as a property of the process itself. Preferred control structures to drive the system correctly are then deduced.

Three different controllability indexes are used [12]: the Relative Gain Array (RGA), the Condition Number (CN) and the Morari Resiliency Index (MRI). The RGA index is used to determine the interaction among control loops in a multivariable process. It is defined as the ratio of the open-loop gain for a selected output when all the other loops of the process are open, to its open-loop gain when all the other loops are closed. RGA of a complex non-singular matrix $M$ is calculated as indicated in Eq. (10), where $\times$ denotes element by element multiplication (Hadamard product).

$$RGA (M) = M \times (M^{-1})$$ (10)

Pairings showing RGA close to unity matrix at frequencies around bandwidth are preferred. This rule favors minimal interaction between loops, which means
independence of the loops. Being the loops independent, stability problems caused by
interaction are prevented. Numbers around 0.5 indicate relevant interaction. The RGA
indicates other useful control properties. One of the most important is that structures
with large RGA elements around the bandwidth frequency are difficult to control due to
sensitivity to input uncertainty [13].

The second controllability index analyzed is the condition number (CN). It also
proceeds from the singular value decomposition of the transfer function. The CN of a
matrix is defined as the ratio between the maximum and minimum singular values and it
is typically used for the selection of the control structure. It provides a numerical
indication of the sensitivity balance in a multivariable system. Large condition numbers
indicate unbalanced sensitivity and also sensitivity to changes in process parameters,
Therefore, structures with small CNs are preferred.

Finally, the Morari Resilience Index (MRI) is the minimum singular value of the open-
loop transfer function, [12]. The MRI is a useful measure for determining whether
acceptable control can be achieved. If it is less than one then poor control performances
are expected. Large MRIs over the frequency range of interest are preferred. Usually,
only the steady state value of these controllability indexes is regarded. However, their
analysis in the frequency domain is important.

4. Study of the static behaviour of the system

To test the nonlinearity of the model we obtained steady state values of the reactor
output flow rates $F_{H2}$, $F_{CO}$ and $F_{CO2}$, following changes in ethanol input flow rate
ΔF_{C2H5OH} and in water input flow rate ΔF_{H2O} with respect to a nominal operating point, see OP1 in Table 1. In Figs. 2 to 5 the H$_2$ yield, $\eta_{H2}$ as defined in Eq. (11), ethanol conversion, $x_{C2H5OH}$ defined as shown in Eq. (12), and acetaldehyde conversion, $x_{C2H4O}$, see Eq. (13), are plotted against the changes performed in F$_{H2}$, F$_{CO}$, F$_{CO2}$ by applying successive steps of +10%. The results have been represented according to the following input increases:

- $\Delta F_{C2H5OH}$: 10% Ethanol input increase while keeping the nominal water input flow rate ($F_{H2O}=0.008210$ mol/s).
- $\Delta F_{H2O}$: 10% Water input increase while keeping the nominal ethanol input flow rate ($F_{C2H5OH}=0.001337$ mol/s).
- $\Delta F_{H2O}$ and $\Delta F_{C2H5OH}$: Both water input and ethanol inputs simultaneously 10% increases in both.

The hydrogen yield $\eta_{H2}$ is used to measure the effectiveness of the reforming process. It is calculated as the ratio between the hydrogen molar flowrate at the outlet of stage 3 and the theoretical hydrogen production:

$$\eta_{H2} = \frac{F}{\frac{6}{F_{C2H5OH,L}}}$$

Ethanol and Acetaldehyde conversions, in stage 1 ($x_{C2H5OH}$) and stage 2 ($x_{C2H4O}$) are calculated through the following definitions:

$$x_{C2H5OH} = 1 - \frac{F_{C2H5OH,L}}{F_{C2H5OH,e}}$$

$$x_{C2H4O} = 1 - \frac{F_{C2H4O,L}}{F_{C2H4O,e}}$$
Fig. 2 shows that the static gains in $F_{\text{H}_2}$ are almost linear positive for $F_{\text{C}_2\text{H}_5\text{OH}}$ and for simultaneous $F_{\text{H}_2\text{O}}$ and $F_{\text{C}_2\text{H}_5\text{OH}}$ increases. However, the static gain is negative when only $\Delta F_{\text{H}_2\text{O}}$ is performed. Regarding the hydrogen yield, the slope observed is steeper for perturbations in the input water flow rate than in the ethanol input flow rate. This behaviour is ascribed to higher drops in residence times (note the feed ratio $\text{H}_2\text{O}:\text{C}_2\text{H}_5\text{OH}$ of 6:1 in OP1).

Fig. 3 shows that the behavior of the $F_{\text{CO}}$ is strongly nonlinear with $\Delta F_{\text{C}_2\text{H}_5\text{OH}}$, while $F_{\text{CO}}$ is almost constant when $F_{\text{H}_2\text{O}}$ and $F_{\text{C}_2\text{H}_5\text{OH}}$ increase simultaneously. The negative slope of $F_{\text{CO}}$ with $\Delta F_{\text{H}_2\text{O}}$ is due to a greater extent of the water-gas shift reaction.

Carbon dioxide flow rate behaviour for perturbations in ethanol and water inputs is reported in Fig. 4. As also shown for CO, CO$_2$ flow rates increase due to higher partial pressure as $F_{\text{C}_2\text{H}_5\text{OH},e}$ increases. Conversely, as the residence times decreases for higher water inputs, drops of $F_{\text{CO}_2}$ are observed.

Fig. 5 reports ethanol and acetaldehyde conversion trends for the same variations in the inlet flows as reported above. Minimum ethanol and acetaldehyde conversions of ca. 90% are required for an appropriate closure of the economical balance of the system, as no recycling of unreacted products is planned. As observed in Fig. 5, higher drops in conversion are verified following increases in space velocities. In contrast, the variations in the ethanol feed fulfil the restriction of minimum conversion requirement.
Table 1 summarizes the performance of the reforming unit at different operating points; results were obtained by using the detailed non linear model described in Eq. (7) to (9). When the reactor is fed with a higher flow rate of water, OP$_3$ and OP$_4$, the H$_2$ yield diminishes. The advantage, however, is a marked decrease of the CO molar fraction output, $y_{CO, L}$. We can also observe that it is possible to improve significantly the conversion of stage 1, $x_{C_2H_5OH}$, by increasing the operating temperature, OP$_9$. In this case a relative good yield is obtained, although the molar fraction $y_{CO, L}$ also increases.

We observe that in OP$_5$, with a large $\Delta F_{C_2H_5OH}$ from OP$_1$, the hydrogen yield diminishes and $y_{CO}$ increases significantly. The best performance, with respect to the production of hydrogen ($F_{H_2, L}$) corresponds to the operation points OP$_2$, OP$_4$, and OP$_5$. Among these, OP$_4$ is preferred as less CO is generated, in agreement with PEMFC feed restrictions.

5. **Controllability Analysis**

5.1 Analysis of Linearized Models

Linear models have been obtained by linearising the complete nonlinear model around the different operating points. To derive the linear models, SIMULINK$^\text{®}$ has been employed because it is integrated in MATLAB which provides an extensive range of tools. Unfortunately the accurate linearization tool supplies a state space matrix of the system with a very large dimension. Thus, a model reduction strategy has been applied to transform the original model into a simplified form with lower order while preserving the dynamic characteristics of the original high-order system [14].
Within the Model Order Reduction (MOR) algorithms for Linear Time-Invariant systems, supplied by MATLAB, the following methods can be distinguished: Balanced MOR based on Hankel Norm Approximation, Proper Orthogonal Decomposition, and MOR algorithms based on Kriol subspace methods.

The Balanced MOR technique was used here to achieve such task directly from the 198th order model. Balanced truncation of the 198th order model was achieved by using the function "balred" of the MATLAB Robust Control System Toolbox. It computes a kth order model given by:

\[ G_m = C_m (Is + A_m)^{-1} B_m + D_m \]  \hspace{1cm} (14)

of a possibly non-minimal and not necessarily stable nth order system described by

\[ G = C (Is + A)^{-1} B + D \]  \hspace{1cm} (15)

\[ \| G(jw) - G_m(jw) \|_\infty \leq totbnd, \]  \hspace{1cm} (16)

\[ totbnd = 2 \sum_{i=k+1}^{n} svh(i) \]  \hspace{1cm} (17)

Where k < n, \( \| \cdot \|_\infty \) denotes the infinity norm of the difference between the full order and the truncated models, totbnd is the total error, and svh(i) denotes Hankel Singular Values of the full order model. Hankel Singular Values are defined as the square-roots of the eigenvalues of the system’s reachability and observability grammians. They are considered a measure of the energy of each state in the system and an indication of its contribution to system characteristics. Thus, keeping only the states that contribute more to the system dynamics will reduce the order of the model while preserving most of its characteristics in terms of stability, time response and frequency response. Fig. 6 shows Hankel Singular Values of the 198th order model and the detail of the first 20. We have kept here 10 states in order to preserve the characteristics of the full order model.
Fig. 7 compares the frequency response gains from the 198th order model and those from the reduced 10th order model. The outputs of the two models overlap quite well, leading to the conclusion that the 10th order reduced model is a good approximation of the linearized 198th order model.

5.2 Input-Output Pairing

One of the most common approaches to control a Multi-Input Multi-Output (MIMO) system is to use a diagonal controller, which is often referred to as a decentralized controller. The decentralized control works well if the system is close to diagonal, which means that the plant can be considered as a collection of individual single-input single-output (SISO) sub-plants with no interaction among them. In this case, the controller for each sub-plant can be designed independently. If the interaction between loops is large, then the performance of the decentralized controller may be poor.

Due to the behavior near the operating points the preferred pairing choices seems to be the $F_{\text{C}_2\text{H}_5\text{OH},e} \rightarrow F_{\text{H}_2\text{L}}$ pair and the $F_{\text{H}_2\text{O},e} \rightarrow F_{\text{CO},L}$ pair. This pairing choice is also confirmed by the relative gain array (RGA) matrix of $G_m$. The frequency range of interest is given by the bandwidth frequency, normally defined as the frequency up to which control is effective [8]. The range of frequencies analysed in this work is $10^{-3}-10^0$ rad s$^{-1}$ because it is assumed that the bandwidth will be within this range. In Table 2 we show RGA at zero frequency for all operating points. Fig. 8 shows RGA$_{11}$ values for all the different operating points over the frequency range of interest.
The RGA can also be used to assess the loop interactions. Large off-diagonal elements of the RGA matrix indicate large loop interactions. A plot of the magnitude difference between the diagonal and off-diagonal elements of the RGA matrices is shown in Fig. 9 for OP$_1$, OP$_3$ and OP$_9$. This figure shows that the interactions increase at high frequencies.

From steady state (s=0) to frequencies near $\omega=0.1$ rad/s, the RGA$_{11}$ element is close to 1.26 (see table 2 or Fig. 8), indicating that the diagonal pairing suggested before could be used. However, at higher frequencies the off-diagonal RGA elements decrease to values around 0.6, thus it is better to use a decentralized control for these frequencies. These conclusions are reinforced by observing that in Fig. 8 the three operating points show a similar positive value in the frequency range $10^{-3}$–$10^{-1}$ [rad s$^{-1}$], hence we can expect a good performance for the decentralized controller only for this frequency range.

In Figs. 10 and 11 the CN and the MRI of the properly scaled 2 input x 2 output system are plotted, respectively. As mentioned before, the CN can be used as a controllability measure: If the CN is small the effects of process model mismatch are not likely to be important; conversely, for large CN there may be sensitivity to model mismatches. The best controllability is obtained for OP$_3$, although other operating points, as OP$_4$, OP$_6$, and OP$_9$, show a reasonable controllability. This suggests that increasing $F_{\text{H}_2\text{O}}$ can help to improve the performance of the control. The larger the value of MRI, the more controllable the process is. The lower values observed in Fig. 11 may indicate poor performance.
From the study of the static behaviour of the system we can derive several considerations about the design of a control strategy for the reformer. From the analysis of Fig. 2, we could expect that, for the control of the H₂ flow based on the manipulation of ΔF₂H₅OH, a linear controller should have a good performance. However, such a controller would give an excessive CO flow rate when the operation is far from the nominal operating point, as shown in Fig. 3, and hence leading to an inappropriate performance of the reformer. If we desire to control F₂H₂ and F₂CO at the same time, we could expect that a linear decentralized controller based on the manipulation of ΔF₂H₂O and ΔF₂H₅OH would perform appropriately. To improve the efficiency, or the conversion rates (x₂H₅OH and x₂C₂H₄O), more complex (coupled) controllers could be used. An alternative is employing additional control loops based on the temperature as the manipulating variable, which we are currently exploring.

6. Conclusions

The high non linear behaviour of the proposed three-staged reformer has been confirmed by the steady state analysis and it has been deduced that some operating points are more efficient. In particular, those with a larger water input flowrate allow the diminution of CO. The same situation occurs for those operating points with temperatures slightly higher in the ethanol dehydrogenation stage and the acetaldehyde steam reforming stage. The controllability analysis with the RGA, CN and MRI suggests that for frequencies until 0.1 rad/s a decentralized control should present a good performance. Due to the high nonlinearities and the crossed interactions among reactions, a centralized controller, although more complex, should perform better if the
reformer works far from the nominal operating point. This is the case for the envisaged application, H₂ supply for a varying load fuel cell. Future work will address the design of these controllers taking into account both mass and energy balances.

Acknowledgements

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Nomenclature

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
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<tbody>
<tr>
<td>( C )</td>
<td>concentration (mol m(^{-3}))</td>
</tr>
<tr>
<td>( F )</td>
<td>molar flowrate (mol s(^{-1}))</td>
</tr>
<tr>
<td>( N )</td>
<td>number of components (=6)</td>
</tr>
<tr>
<td>( R )</td>
<td>reaction rate (mol m(^{-3}) s(^{-1}))</td>
</tr>
<tr>
<td>( t )</td>
<td>time (s)</td>
</tr>
<tr>
<td>( T )</td>
<td>temperature (K)</td>
</tr>
<tr>
<td>( v )</td>
<td>superficial velocity (m s(^{-1}))</td>
</tr>
<tr>
<td>( x )</td>
<td>conversion</td>
</tr>
<tr>
<td>( z )</td>
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</tr>
<tr>
<td>( \eta^{H_2} )</td>
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<td>( y )</td>
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Subscripts

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<tr>
<td>( g )</td>
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<tr>
<td>( i )</td>
<td>reaction number, ( i=1,\ldots,3 ) (ethanol decomposition, acetaldehyde reforming, water-gas shift, respectively)</td>
</tr>
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</table>
\( j \) component number, \( j = 1, \ldots, 6 \) (\( \text{C}_2\text{H}_5\text{OH}, \text{H}_2\text{O}, \text{C}_2\text{H}_4\text{O}, \text{H}_2, \text{CO}, \text{CO}_2, \) respectively)

\( \text{C}_2\text{H}_5\text{OH} \) relative to ethanol

\( \text{H}_2\text{O} \) relative to water

\( \text{C}_2\text{H}_4\text{O} \) relative to acetaldehyde

\( \text{H}_2 \) relative to hydrogen

\( \text{CO} \) relative to carbon monoxide

\( \text{S}_1 \) stage 1

\( \text{S}_2 \) stage 2

\( \text{S}_3 \) stage 3

\( T \) total

**Greek letters**

\( \Delta H_{298} \) heat of reaction (J mol\(^{-1}\))

\( v_{ji} \) stoichiometric coefficient of component \( j \) in reaction \( i \)

**References**


Figures

Fig. 1 Schematic representation of the three-staged reforming process under study.

Fig. 2 Molar flow (solid lines) and yield of hydrogen (dotted lines) as function of the incremental input in $F_{H2O,e}$ and $F_{C2H5OH,e}$. Model under OP1 operating conditions: isothermal operation at $T_e=648K$, operating pressure of ca. 1 bar, $F_{T,e}=0.09547 \text{ mol s}^{-1}$, $Y_{C2H5OH,e}:Y_{H2O,e}=1:6$. 
**Fig. 3** Molar flow of carbon monoxide as function of the incremental input in $F_{\text{H}_2\text{O},e}$ and $F_{\text{C}_2\text{H}_5\text{OH},e}$. Operating conditions as indicated in Fig. 2.

**Fig. 4** Molar flow of carbon dioxide as function of the incremental input in $F_{\text{H}_2\text{O},e}$ and $F_{\text{C}_2\text{H}_5\text{OH},e}$. Operating conditions as indicated in Fig. 2.

**Fig. 5** Plot of $x_{\text{C}_2\text{H}_5\text{OH}}$ and $x_{\text{C}_2\text{H}_4\text{O}}$ as function of the incremental input in $F_{\text{H}_2\text{O},e}$ and $F_{\text{C}_2\text{H}_5\text{OH},e}$. Operating conditions as indicated in Fig. 2.

**Fig. 6** Hankel Singular Values of the 198th order model and the detail of the first 20 values.

**Fig. 7** Superposition of Bode plots for the reduced 10th order model, in blue, and the linearized 198th order model, in red.

**Fig. 8** Relative Gain Array of the element (1,1) for the different operating points in the frequency range $10^{-3}$–$10^{-1}$ [rad s$^{-1}$].

**Fig. 9** Difference between diagonal and off-diagonal elements of the Relative Gain Array for three operating points in the frequency range $10^{-3}$–$10^{-1}$ [rad s$^{-1}$].

**Fig. 10** Condition Number behavior at the different operating points in the frequency range $10^{-3}$–$10^{-1}$ [rad s$^{-1}$].
Fig. 11 Morari Resiliency Index at the different operating points in the frequency range $10^{-3} - 10^{-1}$ [rad s$^{-1}$].