



Article Optimization of Methanol Synthesis under Forced Periodic Operation

Carsten Seidel ^{1,*}, Daliborka Nikolić ², Matthias Felischak ³, Menka Petkovska ⁴, Andreas Seidel-Morgenstern ^{3,5} and Achim Kienle ^{1,3}

- ¹ Institute for Automation Engineering, Otto-von-Guericke University, Universit ätsplatz 2, D-39106 Magdeburg, Germany; kienle@mpi-magdeburg.mpg.de
- ² Institute of Chemistry, Technology and Metallurgy, University of Belgrade, Njegoševa 12, 11000 Belgrade, Serbia; daliborkan@ihtm.bg.ac.rs
- ³ Max-Planck-Institut f
 ür Dynamik Komplexer Technischer Systeme Sandtorstrasse 1, D-39106 Magdeburg, Germany; matthias.felischak@ovgu.de
- ⁴ Department of Chemical Engineering, Faculty of Technology and Metallurgy, University of Belgrade Karnegijeva 4, 11000 Belgrade, Serbia; menka@tmf.bg.ac.rs (M.P.); anseidel@ovgu.de (A.S.-M.)
- ⁵ Institute for Process Engineering, Otto-von-Guericke University, Universit ätsplatz 2, D-39106 Magdeburg, Germany
- * Correspondence: carsten.seidel@ovgu.de

Abstract: Traditionally, methanol is produced in large amounts from synthesis gas with heterogeneous $Cu/ZnO/Al_2O_3$ catalysts under steady state conditions. In this paper, the potential of alternative forced periodic operation modes is studied using numerical optimization. The focus is a well-mixed isothermal reactor with two periodic inputs, namely, CO concentration in the feed and total feed flow rate. Exploiting a detailed kinetic model which also describes the dynamics of the catalyst, a sequential NLP optimization approach is applied to compare optimal steady state solutions with optimal periodic regimes. Periodic solutions are calculated using dynamic optimization with a periodicity constraint. The NLP optimization is embedded in a multi-objective optimization framework to optimize the process with respect to two objective functions and generate the corresponding Pareto fronts. The first objective is the methanol outlet flow rate. The second objective is the methanol yield based on the total carbon in the feed. Additional constraints arising from the complex methanol reaction and the practical limitations are introduced step by step. The results show that significant improvements for both objective functions are possible through periodic forcing of the two inputs considered here.

Keywords: methanol synthesis; forced periodic operation; nonlinear optimization; multi-objective optimization; Pareto front

1. Introduction

Methanol is one of the most important raw materials of the chemical industry. It is used to produce paraffins, olefins, other organic chemicals, and fuels [1]. Commonly, it is produced in large amounts from synthesis gas using heterogeneous catalysts like $Cu/ZnO/Al_2O_3$ under steady state conditions. More recently, there has also been a growing interest in dynamic methanol reactor operation in the context of energy storage and power to methanol processes [2]. In these types of processes, green hydrogen is produced from excess wind or solar energy via electrolysis. It reacts with CO/CO_2 from biogas or waste streams to methanol [3–5]. This will result in a more flexible use of electrical energy from renewable resources and simultaneously in a reduced emission of greenhouse gases to the environment. However, due to the fluctuating availability of the reactants, the reactor will face dynamic varying feeds. The focus of this paper is on improving reactor performance compared to conventional steady state operation. The feed can be provided by renewable



Citation: Seidel, C.; Nikolić, D.; Felischak, M.; Petkovska, M., Seidel-Morgenstern, A.; Kienle, A. Optimization of Methanol Synthesis under Forced Periodic Operation. *Processes* **2021**, *9*, 872. https:// doi.org/10.3390/pr9050872

Academic Editor: Francesco Parrino

Received: 19 April 2021 Accepted: 12 May 2021 Published: 15 May 2021

Publisher's Note: MDPI stays neutral with regard to jurisdictional claims in published maps and institutional affiliations.



Copyright: © 2021 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https:// creativecommons.org/licenses/by/ 4.0/). resources, conventional sources, or a mixture of both. Forced periodic operation can be beneficial if the time averaged reactor performance under forced periodic operation is higher than the corresponding steady state value, which is only possible for nonlinear systems (see, e.g., [6]). In general, the idea is not new and has been discussed since the 1960s (see, e.g., [7-10]). For a recent overview, we refer to [6,11]. For a rigorous evaluation of new forced periodic operation modes, a comparison with optimal steady state conditions is essential. First, work on forced periodic operation of methanol synthesis conducted on some Cu/ZnO and Cu/ZnO/Al₂O₃ catalysts at temperatures between 225 and 270 °C and relatively moderate pressures between 1.97 and 2.93 MPa was done by [12,13]. The results indicate a potential of the methanol synthesis for improvement through periodic operation. However, a rigorous optimization of forced periodic operation and corresponding steady states was not performed due to the experimental focus of this work. Hence, a rigorous evaluation of forced periodic operation of methanol synthesis is lacking. This gap is closed in the present paper. Numerical NLP optimization is applied to compare optimal steady state solutions with optimal periodic regimes. Multi-objective optimization is applied to trace out the Pareto fronts to simultaneously maximize the methanol flow rate and the methanol yield with respect to the total carbon feed. Additional constraints arising in practice are integrated. In the first step, a well-mixed isothermal reactor with multiple periodic inputs is considered, namely, CO partial pressure in the feed and total feed flow rate. The underlying mathematical model is based on the lumped kinetic model presented in [14,15]. It accounts for dynamic changes of the catalyst and shows good agreement with steady state and dynamic experimental data from Vollbrecht [16].

2. Kinetic Model

Methanol synthesis from CO/CO_2 , H_2 over a commercial $Cu/ZnO/Al_2O_3$ catalyst considered in this paper comprises the following three reactions

$$CO + 2H_2 \rightleftharpoons CH_3OH_{\prime}$$
 (1)

$$CO_2 + 3H_2 \rightleftharpoons CH_3OH + H_2O, \tag{2}$$

$$CO_2 + H_2 \rightleftharpoons CO + H_2O.$$
 (3)

The first and the second reaction represent methanol formation by hydrogenation of CO and CO₂. The third reaction is the reverse water–gas shift reaction. For the computations in this paper, we use the simplified reaction kinetics presented in [14,15]. It assumes a Langmuir–Hinshelwood mechanism with three different active surface centers. The resulting expressions for the three reaction rates are:

$$r_{\rm CO} = (1 - \phi) k_1 \left(p_{\rm CO} p_{\rm H_2}^2 - \frac{p_{\rm CH_3 OH}}{K_{\rm P1}} \right) \Theta^{\odot} \Theta^{\otimes 4}, \tag{4}$$

$$r_{\rm CO_2} = \phi^2 k_2 \left(p_{\rm CO_2} p_{\rm H_2}^2 - \frac{1}{K_{\rm P2}} \frac{p_{\rm CH_3 OH} p_{\rm H_2 O}}{p_{\rm H_2}} \right) \Theta^{*^2} \Theta^{\otimes^4}, \tag{5}$$

$$r_{\rm RWGS} = \phi (1 - \phi)^{-1} k_3 \left(p_{\rm CO_2} - \frac{1}{K_{\rm P3}} \frac{p_{\rm CO} p_{\rm H_2O}}{p_{\rm H_2}} \right) \Theta^* \Theta^{\odot}, \tag{6}$$

with the reformulated Arrhenius equation:

$$k_j = A_{k,j} \exp\left(-B_j \left(\frac{T_{\text{ref}}}{T} - 1\right)\right).$$
(7)

The reference temperature used is $T_{ref} = 523.15$ K and equilibrium constants are according to [16,17]:

 $lg(K_{P1}) = 13.8140 + 3784.4/T - 9.2833 \cdot lg(T) + 3.1475 \times 10^{-3}T - 4.2613 \times 10^{-7}T^{2},$ (8) $lg(K_{P2}) = 15.0921 + 1581.7/T - 8.7639 \cdot lg(T) + 2.1105 \times 10^{-3}T - 1.9303 \times 10^{-7}T^{2},$ (9) $lg(K_{P3}) = 1.2777 - 2167/T + 0.5194 \cdot lg(T) - 1.037 \times 10^{-3}T + 2.331 \times 10^{-7}T^{2}.$ (10)

The surface coverages needed in Equations (4)–(6) are:

$$\Theta^{\odot} = \left(1 + K_{\rm CO} p_{\rm CO} + K^{\odot}_{\rm CH_3OH} p_{\rm CH_3OH} + K^{\odot}_{\rm CO_2} p_{\rm CO_2}\right)^{-1},\tag{11}$$

$$\Theta^{\otimes} = \left(1 + \sqrt{K_{\mathrm{H}_2} p_{\mathrm{H}_2}}\right)^{-1},\tag{12}$$

$$\Theta^* = \left(1 + \frac{K_{\text{H}_2\text{O}}K_{\text{O}}}{K_{\text{H}_2}} \frac{p_{\text{H}_2\text{O}}}{p_{\text{H}_2}} + K^*_{\text{CO}_2}p_{\text{CO}_2} + K^*_{\text{CH}_3\text{OH}}p_{\text{CH}_3\text{OH}} + K_{\text{H}_2\text{O}}p_{\text{H}_2\text{O}}\right)^{-1}.$$
 (13)

These three surface centers have certain properties and are important for different reactions:

i : \odot for oxidized surface centers, also assumed as active center for CO hydrogenation;

ii : * for reduced surface centers, also assumed as active center for CO₂ hydrogenation;

iii $: \otimes$ as active surface centers for heterolytic decomposition of hydrogen.

It is further assumed that the oxidized surface centers and the reduced surface centers can be reversibly transformed into each other while keeping the total amount of active centers constant [14,15]. To describe this transformation under dynamic conditions, we introduce the following differential equation:

$$\frac{d\phi}{dt} = k_1^+ \left(y_{\rm CO}(\phi_{max} - \phi) - \frac{1}{K_1} y_{\rm CO_2} \phi \right) + k_2^+ \left(y_{\rm H_2}(\phi_{max} - \phi) - \frac{1}{K_2} y_{\rm H_2O} \phi \right).$$
(14)

A maximum amount of 90 % of reduced centers, i.e., $\phi \le 0.9$, was assumed, corresponding to $\phi_{max} = 0.9$ in the above equation [15]. The equilibrium constants K_1 and K_2 (Equation (14)) can be expressed as a function of the free energies according to

$$K_1 = \frac{k_1^+}{k_1^-} = \exp\left(\frac{-\Delta G_1}{RT}\right),\tag{15}$$

$$K_2 = \frac{k_2^+}{k_2^-} = \exp\left(\frac{-\Delta G_2}{RT}\right).$$
 (16)

Parameters used in this paper are given in Table 1. They have been refitted to the experimental data of Vollbrecht [16] using $\phi \leq 0.9$ in the constraint set of the nonlinear least squares problem instead of shifting ϕ by 0.1 as was done in [14,15] and also expressed in Equation (24) by [18]. Hence, the parameter values in this paper are slightly different from those reported in [14].

 Table 1. Parameter values of the methanol kinetic model.

Parameter	Value	Units
A _{k,CO}	0.00673	mol/kg _{cat} /s/bar ³
B _{CO}	26.4549	-
A_{k,CO_2}	0.0430	mol/kg _{cat} /s/bar ³
B_{CO_2}	1.5308	-
$A_{k,RWGS}$	0.0117	mol/kg _{cat} /s/bar ³
B_{RWGS}	15.6154	-

Parameter	Value	Units
$\sqrt{K_{H_2}}$	1.1064	$1/\sqrt{bar}$
K [*] _{CH2OH}	0	1/bar
K_{H_2O}	0	1/bar
K _O	0	_
K _{CO}	0.1497	1/bar
$K^{\odot}_{CH_2OH}$	0	1/bar
$K_{CO_2}^*$	0.0629	1/bar
$K_{CO}^{\widetilde{O}^2}$	0	1/bar
ΔG_1	$0.3357 imes10^3$	J/mol
ΔG_2	$21.8414 imes 10^3$	J/mol
k_1^+	$7.9174 imes 10^{-3}$	1/s
k_2^+	$0.188 imes10^{-4}$	1/s
ϕ_{max}	0.9	-

Table 1. Cont.

3. Reactor Model

In this paper, a well-mixed, isothermal CSTR with constant pressure and ideal gas phase is considered, corresponding to the lab-scale micro-Berty reactor described in [16]. In the following section, the subscripts i and k describe the species (i.e., CH₃OH, CO₂, CO, H₂, H₂O, N₂, total number $N_k = 6$) and the subscript j describes the reaction (total number $N_r = 3$). The superscript *G* denotes the gas phase and *S* the solid phase. Under the abovementioned assumptions, the model equations follow from the overall material balances of each component i according to

$$\frac{dn_i}{dt} = \frac{d}{dt}(n_i^G + n_i^S) = \dot{n}_0 y_{i,0} - \dot{n} y_i + m_{\text{kat}} \sum_{j=1}^{N_r} v_{i,j} r_j,$$
(17)

with

$$u_i^G = y_i n^G, \tag{18}$$

$$u_i^S = m_{\text{kat}} q_{sat} \Theta_i. \tag{19}$$

Therein, Θ_i is the total coverage of component i at the different surface centers of the solid phase. It is assumed that the Θ_i s are in equilibrium with the gas phase and therefore depend on the partial pressures of the gas phase according to $\Theta_i = \Theta_i(p_1, p_2, ..., p_{N_k})$, as described in detail in Appendix A. Further, due to the assumption of an ideal gas phase and a constant pressure, n^G is constant. Therefore, the time derivatives of n_i^G and n_i^S are

$$\frac{dn_i^G}{dt} = n_G \frac{dy_i}{dt},\tag{20}$$

$$\frac{dn_i^S}{dt} = m_{\text{kat}} q_{\text{sat}} \sum_{k=1}^{N_k} \frac{\partial \Theta_i}{\partial p_k} \frac{dp_k}{dt}.$$
(21)

The derivatives $\partial \Theta_i / \partial p_k$ are also explained in Appendix A. Substituting Equations (20) and (21) into Equation (17) yields with $p_k = py_k$

$$n^{G}\frac{dy_{i}}{dt} + m_{\mathrm{kat}}q_{\mathrm{sat}}p\sum_{k=1}^{N_{k}}\frac{\partial\Theta_{i}}{\partial p_{k}}\frac{dy_{k}}{dt} = \dot{n}_{0}y_{i,0} - \dot{n}y_{i} + m_{\mathrm{kat}}\sum_{j=1}^{N_{r}}v_{i,j}r_{j}$$
(22)

The total material balance is obtained by summation of the component material balances over all species:

$$m_{\rm kat}q_{\rm sat}p\sum_{i=1}^{N_k}\sum_{k=1}^{N_k}\frac{\partial\Theta_i}{\partial p_k}\frac{dy_k}{dt} = \dot{n}_0 - \dot{n} + m_{\rm kat}\sum_{i=1}^{N_k}\sum_{j=1}^{N_r}\nu_{i,j}r_j.$$
(23)

Equation (23) is used for the calculation of the flow rate \dot{n} at the outlet, which is, in general, different from the flow rate at the inlet, because the number of moles and also the volumes are not preserved by chemical reaction network Equations (1)–(3) considered here. For the numerical calculations, the flow rate at the outlet in the component material balances (22) was eliminated using the total material balance (23). Finally, conversion of the material balances to volumetric quantities can be done using the ideal gas law

$$n^{G} = \frac{pV^{G}}{RT}, \qquad \dot{n}_{0} = \frac{p_{0}\dot{V}_{0}}{RT_{0}}, \qquad \dot{n} = \frac{p\dot{V}}{RT},$$
 (24)

with equal pressures and temperatures in the in- and outlet $p_0 = p, T_0 = T$ for the isothermal and isobaric mode of operation considered here.

4. Methods

4.1. Steady State Optimization

For a rigorous evaluation of improvements by forced periodic operation, the calculation of optimal steady state conditions is required as a reference. Steady state optimization is formulated as a nonlinear program (NLP) according to

$$\max_{\mathbf{x},\mathbf{y}} f$$
$$\mathbf{h}(\mathbf{x},\mathbf{y}) = \mathbf{0}$$
$$\mathbf{g}(\mathbf{x},\mathbf{y}) < \mathbf{0}$$
$$\mathbf{lb} \le \mathbf{x} \le \mathbf{ub}$$

where *J* represents a suitable performance criterion (production rate or yield), **y** the state variables and **x** additional optimization variables. In the present paper, pressure, temperature and the steady state flow rate at the inlet are fixed and the inlet composition of the reaction mixture is optimized. Details and bounds on the variables are given in the Results section. Equality constraints are represented by the model equations. Additional equality and inequality constraints are also discussed in the Results section. The NLP problem was solved in MATLAB 2020b using fmincon with a multi-start algorithm with 250 starting points to avoid suboptimal local minima, which are a well-known problem for the present type of nonlinear and hence nonconvex optimization problem (see, e.g., Horst and Tuy [19].

4.2. Optimization of Forced Periodic Operation

Following the ideas in [20–23], simultaneous periodic forcing of two input variables $u_{\text{periodic,1}}$ and $u_{\text{periodic,2}}$ is considered, which is often more promising than periodic forcing of a single input variable. In particular, the focus of this paper is two harmonic forcing functions according to

$$u_{\text{periodic},1}(t) = u_{\text{SS},1}(1 + A_1 \cos(\omega t)),$$
 (25)

$$u_{\text{periodic,2}}(t) = u_{\text{SS,2}}(1 + A_2 \cos(\omega t + \Delta \phi)).$$
(26)

The forcing parameters to be optimized are the amplitudes A_1 , A_2 , the forcing frequency ω and the phase shift $\Delta \phi$ between the two input variables. Additionally, the operating point and the corresponding inputs u_{SS} have to be optimized to find the most promising operating point for the forced periodic operation. In general, this operating point can differ from the optimal steady state optimization. The corresponding optimization problem is formulated as

$$max f$$

$$y(0) = y(\tau)$$

$$M(y)\dot{y} = h(x, y)$$

$$g(x, y) < 0$$

$$lb < x < ub$$

where the first condition implies periodicity of the solution. The length of the period τ is determined by the forcing frequency ω according to $\tau = 2\pi/\omega$. The calculation of periodic solutions requires a solution of the dynamic model equations, as indicated by the dynamic constraints. Further inequality constraints will be discussed in more detail in the Results section. The dynamic optimization problem was solved in MATLAB 2020b with fmincon and a multi-start optimization approach with a sequential approach (see, e.g., Cervantes and Biegler [24]).

4.3. Multi-Objective Optimization

In the present case, multiple, rather than single, objectives are of interest for evaluating the reactor performance. The two objective functions to be considered in this paper are

$$J_1 = \overline{n}_{\rm CH_3OH},\tag{27}$$

$$J_2 = Y_{\text{Carbon}} = \frac{n_{\text{CH}_3\text{OH},out}}{\overline{n}_{\text{Carbon},in}}.$$
(28)

The first objective function is the time averaged molar outlet flow rate of methanol. The second objective function is the time averaged yield of methanol, based on the carbon species entering the reactor. Average values were calculated using a single period with 1000 equidistant time points.

The system for methanol synthesis is complex and nonlinear. A straightforward approach with a weighted sum of both objectives may give poor results for steep or even nonconvex Pareto fronts [25,26]. Therefore, the ϵ -constraint method is applied. This is a scalarization technique which reduces the multi-objective optimization problem to a single objective optimization with the second objective function as an additional constraint [27–29]. The quantity ϵ is used as a parameter to change this constraint between a given upper and lower bound of the second objective. For the optimization of periodic regimes, the multi-objective problem formulation reads, e.g.,

$$\max_{x,y} J_1$$
s.t. $J_2 \leq \epsilon$
 $y(0) = y(\tau)$
 $M(y)\dot{y} = h(x,y)$
 $g(x,y) < 0$
 $lb \leq x \leq ub$

An analogous formulation applies to the multi-objective steady state optimization. Both multi-objective optimization problems are solved again in MATLAB 2020b with fmincon, and a multi-start heuristic with 250 starts per point on the Pareto front.

5. Results

5.1. Steady State Multi-Objective Optimization

For steady state operation pressure, temperature and volumetric flow rate at the inlet are fixed and the composition of the reaction mixture at the inlet is optimized. Parameter values correspond to a lab-scale micro-Berty reactor according to Vollbrecht [16]. They are summarized together with the fixed operational parameters in Table 2.

Parameter	Value	Units
p	60	bar
T	473	К
$y_{N_2,0}$	0.15	-
\dot{V}_0	$1.14 imes10^{-7}$	m ³ /s
V^G	$1.03 imes10^{-4}$	m ³
m _{kat}	0.00395	kg
$q_{ m sat}$	0.98	mol/kg

Table 2. Process parameters.

For steady state optimization, the following additional constraints are used:

$$y_{\rm CO,0} + y_{\rm CO_2,0} + y_{\rm H_2,0} + y_{\rm N_2,0} = 1, \tag{29}$$

$$y_{\rm CO,0} + y_{\rm CO_2,0} \ge 0.01,$$
 (30)

$$\begin{bmatrix} 0\\0\\0.5 \end{bmatrix} \leq \begin{bmatrix} y_{\mathrm{CO},0}\\y_{\mathrm{CO}_{2},0}\\y_{\mathrm{H}_{2},0} \end{bmatrix} \leq \begin{bmatrix} 1\\1\\0.85 \end{bmatrix}.$$
(31)

The first constraint represents the summation condition of all species in the feed. The second constraint guarantees at least 1% of carbon in the feed to ensure methanol production. Furthermore, lower and upper bounds for the feed species are implemented. Hydrogen is assumed to always be available in excess. A multi-objective optimization regarding Equations (27) and (28) was done and the resulting Pareto front is shown in Figure 1. Two reference points OP1 and OP4 are marked on the Pareto front of steady state operation represented by the crosses in Figure 1. At the Pareto optimal steady state operating point OP1, the methanol flow rate is 336 mmol/min/kg_{cat} with a yield of 61%. At the Pareto optimal steady state operating point OP4, the methanol flow rate is 413 mmol/min/kg_{cat} with a yield of 52%. The corresponding optimal steady state feed concentrations are shown in the first three diagrams of Figure 2. It can be seen from Figure 2 that the optimal feed contains more CO than CO₂. This is a well-known fact, which follows from the equilibrium limitations of the reaction network and in particular the water inhibition for methanol production (see, e.g., Vollbrecht [16] and references therein). The optimal CO and CO_2 concentrations in the feed are continuously decreasing from the left to the right along the Pareto front.



Figure 1. Pareto fronts for methanol synthesis: x: optimal steady state operation, \Box : optimal forced periodic operation, \bigcirc : chosen operating points.



Figure 2. Optimized parameter values along the Pareto fronts in Figure 1: x: optimal steady state operation, \Box : optimal forced periodic operation. Subplots (**a**–**c**) show the optimized feed concentration of CO₂, CO and H₂. Subplots (**d**–**h**) show the optimized forcing parameters A_{CO}, A_{N₂}, A_F, period time τ and phase shift $\Delta\phi$.

5.2. Multi-Objective Optimization of Forced Periodic Operation

Using the nonlinear frequency response method [30,31], it was found that simultaneous periodic modulation of the CO feed concentration and the volumetric flow rate at the inlet according to

$$y_{\rm CO,0}(t) = y_{\rm CO,0,SS}(1 + A_{\rm CO}\cos(\omega t)),$$
 (32)

$$\dot{V}_0 = \dot{V}_{0,\text{SS}}(1 + A_F \cos(\omega t + \Delta \phi)). \tag{33}$$

are most promising for forced periodic operation.

To satisfy the summation condition for all $y_{i,0}$ at any time t, the N₂ inert gas feed concentration is also varied periodically in an opposite way in order to compensate the periodic change of the CO feed concentration

$$y_{N_{2},0}(t) = y_{N_{2},0,SS}(1 - A_{N_{2}}\cos(\omega t)).$$
(34)

For $\dot{V}_{0.SS}$, a fixed value of $1.14 \times 10^{-7} \text{ m}^3/\text{s}$, and for $y_{\text{N}_2,0,\text{SS}}$, a fixed value of 0.15 was used (see also Table 2). The other parameter values in Table 2 were also used for the optimization of forced periodic operation. With the CO modulation, the carbon content is varied. The flow rate modulation affects the residence time.

Additional constraints used for the optimization of forced periodic operation are

$$y_{\rm CO,0,SS} + y_{\rm CO_2,0,SS} + y_{\rm H_2,0,SS} + y_{\rm N_2,0,SS} - 1 = 0,$$
(35)

$$A_{\rm CO}y_{\rm CO,0,SS} - A_{\rm N_2}y_{\rm N_2,0,SS} = 0, (36)$$

$$y_{\rm CO,0,SS} + y_{\rm CO_2,0,SS} \ge 0.01,$$
 (37)

$$y_i(0) - y_i(\tau) = 0,$$
 (38)

 $y_{\rm CH_3OH}(0) + y_{\rm CO}(0) + y_{\rm CO_2}(0) + y_{\rm H_2}(0) + y_{\rm H_2O}(0) + y_{\rm N_2}(0) - 1 = 0,$ (39)

$$\begin{bmatrix}
0 \\
0 \\
0 \\
0.5 \\
0 \\
0 \\
0 \\
0 \\
18 \\
-\pi \\
0
\end{bmatrix}
\leq
\begin{bmatrix}
y_{CO,0} \\
y_{CO_2,0} \\
M_F \\
A_F \\
A_F \\
1 \\
1 \\
0.85 \\
1 \\
1 \\
3600 \\
\pi \\
1
\end{bmatrix}
.$$
(40)

Therein, Equations (32) and (34), together with constraints (35) and (36), guarantee that the summation conditions at the inlet are satisfied at each time t. Equation (38) represents the periodicity constraint, with an appropriate summation condition for the state variables at the beginning of the period (39).

Lower and upper bounds for the optimization variables are summarized in Equation (40). The amplitudes are bounded between 0 and 1. The phase shift is bounded between $-\pi$ and π . It was found (data not shown) that the improvement through periodic forcing increases with increasing frequency. However, in view of practical implementation, the period time τ is limited here between 18 s and 3600 s.

An additional constraint arises when taking a look at the reaction Equations (1) –(3). The reactions of methanol synthesis are mole number and volume reducing. Therefore, an additional path constraint for the outlet flow rate is required to ensure a positive outlet flow at constant pressure at any time point according to

$$\dot{V}(t) > 0. \tag{41}$$

Results for the multi-objective optimization of forced periodic optimization are also shown in Figure 1 and compared to optimal steady state operation. Percentages of improvement for the individual objective functions (27) and (28) relative to the optimal steady state operating points OP1 and OP4 are also given in this figure. The improvement in the methanol flow rate is between 15% at OP5 and 24% at OP3 and increases from left to right along the Pareto front. The improvement for the yield is between 17% at OP6 and 5% at OP3 and decreases from left to right along the Pareto front. The corresponding optimized parameter values for forced periodic operation are shown in Figure 2. The trends of the mean inlet concentrations $y_{i,0,SS}$ follow the trends of the corresponding steady state values in the first three diagrams. The amplitude of CO forcing is increasing from left to right, whereas the amplitude of N_2 is decreasing correspondingly as explained above. The amplitude of the flow rate forcing is almost constant. The predicted optimal phase shift between CO concentration and flow rate is low over the whole range. The period τ is close to the lower boundary of 18 s most of the time. The outlier between a yield of 0.55 and 0.6 most likely represents a close by local optimum. The inlet concentrations and forcing parameters for the presented operating points in Figure 1 are summarized in Table 3. Finally, the values of the variables y_i , \dot{V} and ϕ and the relative amount of reduced surface centers, at the beginning of the period at t = 0, are shown in Figure 3 to illustrate the evolution of the periodic solutions along the Pareto front.



Figure 3. Initial values y(0) for forced periodic operation along the Pareto fronts in Figure 1. Subplots (**a**–**f**) show the initial values of CH₃OH, CO₂, CO, H₂, H₂O and N₂. Subplot (**g**) shows the initial values of the relative amount of reduced surface centers ϕ . Subplot (**h**) shows the initial values of the flow rate at the outlet.

ОР	<i>у</i> со ₂ ,0,55	$y_{\rm CO,0,SS}$	$y_{\rm H_2,0,SS}$	A_F	A _{CO}	A_{N_2}	au in s	$\Delta \phi$
OP1	0.0296	0.184	0.6453	-	-	-	-	-
OP2	0.0266	0.152	0.6714	1	0.9866	1	34	0.1637
OP3	0.0176	0.1206	0.7118	0.97	1	0.804	18.5	-0.253
OP4	0.0314	0.273	0.5433	-	-	-	-	-
OP5	0.0345	0.238	0.577	0.9378	0.6291	1	18	-0.04
OP6	0.0266	0.152	0.6714	1	0.9866	1	34	0.1637

 Table 3. Operating conditions for presented operating points from Figure 1.

In summary, we find that the Pareto front of optimal forced periodic operation always lies above the Pareto front of optimal steady state operation, with somewhat substantial improvements for the conditions considered here.

6. Conclusions

In this contribution, we have shown theoretically that the performance of methanol synthesis from synthesis gas with a commercial $Cu/ZnO/Al_2O_3$ catalyst can be improved significantly by periodic forcing of the CO feed concentration and the phase-shifted feed flow rate compared to optimal steady state operation. Improvements were measured in terms of methanol flow rate and methanol yield relative to the total carbon in the feed. The focus was on harmonic forcing functions, and a well-mixed isothermal CSTR. Due to the complexity of the methanol synthesis using a $Cu/ZnO/Al_2O_3$ catalyst, the isothermal CSTR should be considered as a first reasonable step to provide some fundamental insight into the effect of forced periodic operation on methanol synthesis.

In industry, methanol synthesis is usually performed in a cooled fixed bed reactor. Perfect isothermicity cannot be achieved in such a reactor and additional temperature effects play a role. They may put additional constraints on practical reactor operation but also offer additional degrees of freedom for periodic forcing, which we want to investigate in our future work. However, since the modeling of a fixed bed reactor leads to a set of partial differential equations, more advanced numerical methods are required for optimization purposes.

The results in this paper show that excess hydrogen in the feed is beneficial for the methanol flow rate and the yield in terms of total carbon for steady state and forced periodic operation. In practice, the cost of hydrogen can be a major factor, particularly for renewable methanol. For reduced hydrogen feed, the production rate of methanol and the yield will be reduced for steady state as well as forced periodic operation compared to the over-stoichiometric case considered in this paper. However, it is expected that forced periodic operation will be still superior to the steady state operation due to the additional degrees of freedom. To quantify the economic benefit, further calculations are required using an economic objective function accounting for the specific hydrogen supply cost. In view of practical applications, this will be done for a fixed bed reactor in our future work. Further, we are aiming at an experimental validation of the theoretical findings presented in this paper. Experiments, however, are not trivial due to the relatively high pressure and the properties of the reactants considered here and are therefore clearly beyond the scope of the present paper and a subject for future research.

Author Contributions: Conceptualization, C.S. and A.K.; methodology, C.S., A.K., D.N. and M.P.; numerical studies , C.S.; writing original draft , C.S. and A.K.; review and editing, A.K., M.F., A.S.-M., D.N. and M.P.; Funding acquisition, A.S.-M., A.K. and M.P. All authors have read and agreed to the published version of the manuscript.

Funding: Gefördert durch die Deutsche Forschungsgemeinschaft (DFG)—SPP2080 "Katalysatoren und Reaktoren unter dynamischen Betriebsbedingungen für die Energiespeicherung und -wandlung"—Förderkennzeichen Kl 417/6-1, PE 2915/1-1, SE 586/24-1 (Projektnummer: 406561907). Financial support by the German Research Foundation (DFG) is gratefully acknowledged through the priority program SPP 2080 under grant Kl 417/6-1, PE 2915/1-1, SE 586/24-1 (project number: 406561907).

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

Data Availability Statement: Not applicable.

Conflicts of Interest: The authors declare no conflict of interest.

Nomenclature

A_{x}	amplitude of input CO, N_2 or volumetric flow rate
$\Delta G [J/mol]$	Gibbs free energy
J	objective function
$k_i [\mathrm{mol/s/kg_{cat}/bar^3}]$	reaction rate constant for $j = 1, 2$
$k_3 [\text{mol/s/kg}_{cat}/\text{bar}]$	reaction rate constant for $j = 3$
k_1^+, k_2^+ [1/s]	reaction rate constant for oxidation and reduction of the catalyst
K_i	adsorption constant
$K_{Pi}[1/bar^2]$	equilibrium constant for $j = 1, 2$
$K_{P3}[-]$	equilibrium constant for $j = 3$
K_1, K_2	equilibrium constant for oxidation and reduction of the catalyst
\dot{n} [mol/s]	molar flow rate
$\dot{n}_i [\text{mol/s}]$	molar flow rate of component <i>i</i>
<i>p</i> [bar]	pressure
$p_i[\text{bar}]$	partial pressure of compenent <i>i</i>
q_{sat} [mol/kg _{cat}]	specific number of surface centers
R [J/K/mol]	gas constant
$r_i [mol/s/kg_{cat}]$	rate of reaction <i>i</i>
T[K]	temperature
- [] U	general input
$V_C [m^3]$	volume of the gas phase in the reactor
$\dot{V} [m^3/s]$	volumetric flow rate
x	optimization variables
v	general output variables
y Vi	mole fraction of component <i>i</i>
YCarbon	vield of methanol based on total carbon in the feed
Greek letters	,
Θ	relative number of free surface centers
τ [s]	period time
φ _{max}	maximum fraction of reduced centers on the catalyst surface
φ	fraction of reduced centers on the catalyst surface
$\Delta \phi$	phase shift
$\omega[1/s]$	frequency
Subscripts	1 5
periodic	forced periodic feed stream
SS	steady state
0	feed stream
i	component ($i = 1 \text{ CH}_3\text{OH}$, $i = 2 \text{ CO}_2$, $i = 3 \text{ CO}$, $i = 4 \text{ H}_2$,
	$i = 5 \text{ H}_2\text{O}, i = 6 \text{ N}_2$
i	reaction ($j = 1$ CO hydrogenation, $j = 2$ CO ₂ hydrogenation,
)	i = 3 RWGS
Superscripts	
G	gas phase
S	solid phase
*	reduced surface center
\odot	oxidized surface center
\otimes	reduced surface center
Abbreviations	
OP	Operating point

Appendix A. Partial Derivatives of the Isotherms of the Adsorbed Species

For the reactor model, the partial derivatives of the isotherms of the adsorbed species $\frac{\partial \Theta_i}{\partial p_k}$ are required. Therein, Θ_i is the total coverage of component i at the different surface centers of the solid phase. According to Equations (11)–(13), the following components are adsorbed at the different surface centers:

- (i) \odot oxidized surface centers: CH₃OH, CO₂, CO
- (ii) * reduced surface centers: CH_3OH , CO_2 , H_2 , H_2O
- (iii) \otimes surface centers for heterolytic decomposition of hydrogen: H₂

The corresponding adsorption isotherms are as follows:

$$\Theta_i^{\odot} = K_i^{\odot} p_i \Theta^{\odot}$$
, for i=CH₃OH, CO₂, CO (A1)

$$\Theta_i^* = K_i^* p_i \Theta^*, \quad \text{for i=CH_3OH, CO_2, H_2O}$$
(A2)

$$\Theta_H^* = \sqrt{K_{H_2} p_{H_2} \Theta^*}, \quad \text{and} \tag{A3}$$

$$\Theta_{H}^{\otimes} = \sqrt{K_{H_2} p_{H_2}} \Theta^{\otimes} \tag{A4}$$

with Θ^{\odot} , Θ^* , Θ^{\otimes} according to Equations (11)–(13). According to the reaction mechanism presented in [14,16], it is assumed that hydrogen, i.e., H_2 from the gas phase, is adsorbed and simultaneously decomposed into elementary hydrogen, so that in the solid phase only elementary hydrogen is present. This explains the square root in Equations (A3) and (A4).

The total coverage of component i in the solid phase is the sum over all surface centers which adsorb this component, for methanol this is, for example

$$\Theta_{CH_3OH} = \Theta_{CH_3OH}^{\odot} + \Theta_{CH_3OH}^* \tag{A5}$$

Accordingly, the derivatives are also obtained by summation of the corresponding individual derivatives, e.g.,

$$\frac{\partial \Theta_{CH_3OH}}{\partial p_k} = \frac{\partial \Theta_{CH_3OH}^{\odot}}{\partial p_k} + \frac{\partial \Theta_{CH_3OH}^{*}}{\partial p_k}, \quad \text{for all } k$$
(A6)

Finally, it should be noted that the set of equations is given here in full, despite the fact that some of the parameters are zero (see Table 1).

References

- 1. Asinger, F. Methanol—Chemie und Energierohstoff; Springer: Berlin/Heidelberg, Germany, 1986.
- Kalz, K.F.; Kraehnert, R.; Dvoyashkin, M.; Dittmeyer, R.; Gläser, R.; Krewer, U.; Reuter, K.; Grunwaldt, J.D. Future challenges in heterogeneous catalysis: Understanding catalysts under dynamic reaction conditions. *ChemCatChem* 2017, 9, 17–29. [CrossRef]
- Olah, G.A. Beyond oil and gas: The methanol economy. *Angew. Chem.* 2005, 44, 2636–2639. [CrossRef]
 Martín, M. Methodology for solar and wind energy chemical storage facilities design under uncertainty: Methanol production
- from CO₂ and hydrogen. *Comput. Chem. Eng.* **2016**, 92, 43–54. [CrossRef]
- Raeuchle, K.; Plass, L.; Wernicke, H.J.; Bertau, M. Methanol for renewable energy storage and utilization. *Energy Technol.* 2016, 4, 193–200. [CrossRef]
- 6. Petkovska, M.; Nikolić, D.; Seidel-Morgenstern, A. Nonlinear frequency response method for evaluating forced periodic operations of chemical reactors. *Isr. J. Chem.* **2018**, *58*, 663–681. [CrossRef]
- 7. Douglas, J.M.; Rippin, D.W.T. Unsteady state process operation. Chem. Eng. Sci. 1966, 21, 305–315. [CrossRef]
- 8. Horn, F.J.M. Periodic countercurrent processes. Ind. Eng. Chem. Process Des. Dev. 1967, 6, 20–30. [CrossRef]
- 9. Baily, J.E. Periodic operation of chemical reactors: A review. Chem. Eng. Commun. 1973, 1, 111–124. [CrossRef]
- 10. Silveston, P.L. Periodic operation of chemical reactors—A review of the experimental literature. *Sadhana* **1987**, *10*, 217–246. [CrossRef]
- 11. Silveston, P.L.; Hudgins, R.R. Periodic Operation of Reactors; Elsevier: Amsterdam, The Netherlands, 2013.
- 12. Nappi, A.O.; Fabbricino, L.; Hudgins, R.R.; Silveston, P.L. Influence of forced feed composition cycling on catalytic methanol synthesis. *Can. J. Chem. Eng.* **1994**, 72, 657–671. [CrossRef]
- 13. Chanchlani, K.G.; Hudgins, R.R.; Silveston, P.L. Methanol synthesis under periodic operation: An experimental investigation. *Can. J. Chem. Eng.* **1985**, *63*, 963–970. [CrossRef]

- 14. Seidel, C.; Jörke, A.; Vollbrecht, B.; Seidel-Morgenstern, A.; Kienle, A. Kinetic modelling of methanol synthesis from renweable resources. *Chem. Eng. Sci.* 2018, 175, 130–138. [CrossRef]
- 15. Seidel, C.; Jörke, A.; Vollbrecht, B.; Seidel-Morgenstern, A.; Kienle, A. Corrigendum to 'Kinetic modeling of methanol synthesis from renewable resources' (Chem. Eng. Sci. 175 (2018) 130-138) . *Chem. Eng. Sci.* 2020, 223, 115724. [CrossRef]
- Vollbrecht, B. Zur Kinetik der Methanolsynthese an einem technischen Cu/ZnO/Al₂O₃-Katalysator. Ph.D. Thesis, Otto-von-Guericke Universität Magdeburg, Germany, 2007.
- 17. Skrzypek, J.; Sloczyński, J.; Ledakowicz, S. *Methanol Synthesis: Science and Engineering*; Polish Scientific Publishers: Warsaw, Poland , 1994.
- Slotboom, Y.; Bos, M.J.; Pieper, J.; Vrieswijk, V.; Likozar, B.; Kersten, S.R.; Brilman, D. Critical assessment of steady-state kinetic models for the synthesis of methanol over an industrial Cu/ZnO/Al₂O₃ catalyst. *Chem. Eng. J.* 2020, 389. [CrossRef]
- 19. Horst, R.; Tuy, H. Global Optimization; Springer: Berlin, Germany, 1993.
- Nikolić, D.; Seidel-Morgenstern, A.; Petkovska, M. Nonlinear frequency response analysis of forced periodic operation of non-isothermal CSTR with simultaneous modulation of inlet concentration and inlet temperature. *Chem. Eng. Sci.* 2015, 137, 40–58. [CrossRef]
- 21. Nikolić, D.; Seidel-Morgenstern, A.; Petkovska, M. Periodic operation with modulation of inlet concentration and flow rate. Part I: Nonisothermal continuous stirred-tank reactor. *Chem. Eng. Technol.* **2016**, *39*, 2020–2028. [CrossRef]
- 22. Nikolić, D.; Seidel-Morgenstern, A.; Petkovska, M. Periodic operation with modulation of inlet concentration and flow rate. Part II: Adiabatic continuous stirred-tank reactor. *Chem. Eng. Technol.* **2016**, *39*, 2126–2134. [CrossRef]
- Felischak, M.; Kaps, L.; Hamel, C.; Nikolić, D.; Petkovska, M.; Seidel-Morgenstern, A. Analysis and experimental demonstration of forced periodic operation of an adiabatic stirred tank reactor: Simultaneous modulation of inlet concentration and total flow-rate. *Chem. Eng. J.* 2021, *410*, 128197. [CrossRef]
- 24. Cervantes, A.; Biegler, L.T. Optimization Strategies for Dynamic Systems. In *Encyclopedia of Optimization*; Springer: Boston, MA, USA. 2008; pp. 2847–2858. [CrossRef]
- 25. Das, I.; Dennis, J.E. A closer look at drawbacks of minimizing weighted sums of objectives for Pareto set generation in multicriteria optimization problems. *Struct. Optim.* **1997**, *14*, 63–69. [CrossRef]
- 26. Marler, R.T.; Arora, J.S. The weighted sum method for multi-objective optimization: New insights. In *Structural and Multidisciplinary Optimization*; Springer: Berlin/Heidelberg, Germany, 2010; doi:10.1007/s00158-009-0460-7. [CrossRef]
- 27. Haimes, Y.V. On a bicriterion formation of the problems of integrated system identification and system optimization. *IEEE Transact. Syst. Man Cybern.* **1971**. [CrossRef]
- 28. Ehrgott, M. *Multicriteria Optimization: Second Edition;* Springer Science & Media: Berlin/Heidelberg, Germany, 2005; pp. 1–323. [CrossRef]
- Maußner, J. Single- and Multi-Objective Reactor Design Under Uncertainty. Ph.D. Theis, Friedrich-Alexander-Universität Erlangen-Nürnberg (FAU), Erlangen, Germany, 2020.
- 30. Nikolić, D.; Seidel, C.; Felischak, M.; Milicić, T.; Kienle, A.; Seidel-Morgenstern, A.; Petkovska, M. Forced periodic operations of a chemical reactor for methanol synthesis—The search for the best scenario based on the nonlinear frequency response method. Part I Single input modulations. 2021. In preparation.
- 31. Nikolić, D.; Seidel, C.; Felischak, M.; Milicić, T.; Kienle, A.; Seidel-Morgenstern, A.; Petkovska, M. Forced periodic operations of a chemical reactor for methanol synthesis—The search for the best scenario based on the nonlinear frequency response method. Part II Simultaneous modulation of two inputs. 2021. In preparation.