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Influence of flow rate, temperature and pressure on multiphase flows of supercritical carbon dioxide and water using multivariate partial least square regression

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Abstract
Supercritical carbon dioxide (scCO₂) is often used to replace harmful solvents and can dissolve a wide range of organic compounds. With a favorable critical point at 31 °C and 7.4 MPa, reaching above the critical point for scCO₂ is fairly accessible. Because of the compressible nature of scCO₂ and the large changes of viscosity and density with temperature and pressure, there is a need to determine the behavior of scCO₂ in microfluidic systems. Here, the influence of how parameters such as flow rate, temperature, pressure, and flow ratio affects the length of parallel flow of water and scCO₂ and the length of the created CO₂ segments are investigated and modeled using multivariate data analysis for a 10 mm long double-y channel. The parallel length and segment size were observed in the laminar regime around and above the critical point of CO₂. The flow ratio between the two fluids together with the flow rate influenced both the parallel length and the segment sizes, and a higher pressure resulted in shorter parallel lengths. Regarding the segment length of CO₂, longer segments were a result of a higher Weber number for H₂O together with a higher temperature in the channel.

Keywords: supercritical fluids, microfluidics, carbon dioxide, partial least square regression, principal component analysis, fluid dynamics, multiphase flow

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(Some figures may appear in colour only in the online journal)
while the liquid-like density will provide sufficient solute-solvent interaction. Hence, chemical processes like synthesis and extraction can be tuned by changing the pressure and temperature. SC-CO₂ can dissolve a wide range of organic compounds due to its low dipole moment, combined with Lewis acid-base interactions. Further, to configure the solubility profile, polar co-solvents such as ethanol or methanol can be added [5].

With a favorable critical point at 31 °C and 7.4 MPa, reaching above the critical point for SC-CO₂ is fairly accessible. However, because of the compressible nature of SC-CO₂ and the large changes of viscosity and density with temperature and pressure, there is a need to determine the behavior of SC-CO₂ in microfluidic systems, especially when there are multiple phases in the system, like H₂O and SC-CO₂. A typical pressure and temperature for a process is 15 MPa and 80 °C, giving a viscosity ratio of approximately 10:1 [4, 6] for H₂O and CO₂. The density ratio between the two phases is approximately 2:1 [3, 6]. To provide a stable control of multiphase systems where the two phases have radically different properties when it comes to viscosity and compressibility has been identified as one of the future challenges in microfluidics by Elvira et al [7]. Computational fluid dynamics (CFD) is an important tool in research as well for industrial applications, and to experimentally characterize the behavior of this type of two phase systems gives data that can be useful for validation of computational models [8, 9]. After a junction where two phases joins, depending on the system parameters, the result can either be a segmented flow, or parallel flow between the two phases. Depending on the application, one or the other is preferred [10].

As stated in [11], fluids in nature, biotechnology and chemistry are rarely single-phase. In the past, microfabrication has been demonstrated using liquid–liquid laminar flow multiphase systems [12, 13], and methods for controlling and different applications of both droplets and interfaces are summarized in [14]. As things stand, most presented multiphase SC-CO₂-microsystems utilize segmented flows, e.g. to extract vanillin from a water solution using a glass-silicon microdevice operating at 39.8 °C and 8.1 MPa to 11.1 MPa [15], and to extract different aromatic monomers from lignin oxidation products [16]. Tiggelaar et al studied the formation of carbamic acid from N-benzylmethylamine and CO₂, and reaction products were observed at pressures up to 30 MPa and a temperature of 100 °C inside a glass-silicon chip [17]. Marre et al have demonstrated a modular glass-silicon-based chip, capable of even higher temperatures and pressures, but here using segmented SC-H₂O multiphase systems. The fluid was heated in a confined section of the chip to 400 °C at 25 MPa while keeping the connections closer to room temperature [18]. Other applications for this chip have been quantum dot synthesis using supercritical hexane [19] and as a pressure-volume-temperature cell for SC-CO₂ where the transition from single phase to multiphase was studied [20].

There has been less studies utilizing and investigating parallel flows of SC-CO₂, although the potential use in extraction and analysis [10]. Ohashi et al studied extraction of tris(acetylacetonato)-cobalt(III) from SC-CO₂ into water and achieved a parallel flow for 0.5 s using a hydrophobic coating and having a guide in between the flows [21]. The experiment was performed at 10 MPa and 45 °C. However, only limited details of the design is mentioned by the authors.

Ogden et al characterized SC-CO₂-water flows for different flow rates and relative ratio of SC-CO₂ and water [22]. In all experiments, the pressure was set to 10 MPa and the temperature to 50 °C. Segmented flow was mostly observed, although parallel flow was observed for high ratios of SC-CO₂ relative to water. At high flow rates the parallel flow was observed to be instable and was denoted wavy.

Multivariate data analysis is widely used in chemistry [23] and for pharmaceutical applications [24], as well as being a valuable tool in the processing industry [25]. In microsystems, and especially in microfluidics, multivariate techniques have mainly been used to calibrate different responses to spectroscopic data generated from Raman spectroscopy [26–28] and x-ray fluorescence spectroscopy [29], or to analyze responses from microsensors or microsensor arrays [30, 31]. Here we demonstrate how multivariate data analysis can be used to model and predict segment length and parallel length in the complex multiphase system of SC-CO₂ and H₂O.

2. Theory

For microfluidic systems, dimensionless numbers are often used to characterize the properties of the flow, and for small systems where gravity is negligible, the Reynolds number (Re), capillary number (Ca) and Weber number (We) are the most interesting [32].

For our experiments, Re, Ca, and We were calculated at 10 MPa and the temperature to 50 °C. Segmented flow was mostly observed, although parallel flow was observed for high ratios of SC-CO₂ relative to water. At high flow rates the parallel flow was observed to be instable and was denoted wavy.

Guillot and Colin studied water and oils with varying viscosities flowing in parallel [33]. Three different behaviors were observed; parallel, parallel broken up to segmented droplet flow, and droplets formed at the T-junction, the latter not having reproducible droplet sizes. For the parallel flow breaking up in the channel, they found that the stability of the
flow is linked to the confinement of the flow and cannot be described in terms of a competition between viscous forces and the capillary forces, i.e. Ca, but is determined by the channel wetting conditions, geometry and the viscosity.

Garstecki et al studied the mechanism of droplets breaking up in a T-junction for low Re of two-phase flows of silicon oil and water, and water and nitrogen gas [34]. The dominant effect of breakup was found to be the pressure-drop across the emerging segments, where the size of the segments were dependent on the viscosity of the continuous phase, the geometry of the device and the interfacial tension between the two phases.

In a review by Hessel et al different gas-liquid continuous phase micro reactors are presented [35]. For dispersed-phase micro reactors, the gas-liquid flow ratio is reported to be a key parameter to obtain annular flows, for which at high values, a thin liquid annulus film surrounds a gas core. For these micro reactors, there is generally a degree of uncertainty concerning the fluid geometries and available interfacial areas, and flow disturbances can for these systems cause entrainment of the liquid film into the gas core.

Working with fluids with varying density, the volumetric flow rate will be different inside the pump compared to the chip. By using conservation of mass, the volumetric flow rate can be modified to correct for density variations along the channel according to

$$\rho_{\text{pump}} Q_{\text{pump}} = \rho_{\text{chip}} Q_{\text{chip}},$$

where $Q$ is the volumetric flow rate and $\rho$ is the density of the fluid. Thus, the fluid velocity on the chip, accounting for the preservation of mass, in a heated channel can be described as

$$U = \frac{Q_{\text{pump}} \rho_{\text{pump}}}{A \rho_{\text{chip}}},$$

where $A$ is the fluid cross-section area in the channel. Hence, the flow velocity is affected by changes in the pressure and temperature.

With multivariate data analysis it is possible to detect and model the underlying factors affecting a multivariate system. In principal component analysis (PCA), the orthogonal directions of maximum variation, i.e. the principal components (PC), are found. A score plot is generated by projecting the data, here variables and responses, on the PCs showing general trends and the plot can also be used for data classification [36]. The corresponding loading plot visualizes the relation between the variables. Partial least squares regression (PLS) is a similar procedure, but here, the covariance between the variables and the responses are maximized in order to find a linear regression model describing the system [37].

3. Experimental

3.1. Manufacturing

The chip used had a double-y channel with a 10 mm long channel, figure 2, etched from one 4″ Borofloat-33 glass wafer (Schott), the channel wafer, bonded to another Borofloat-33 wafer, figure 3, only with trenches for connections etched. After etching, the channel was 40 μm in height, and 193 μm in width.

Both wafers were initially co-processed, with the second shallow etching for channels only done on the channel wafer. Processed in a cleanroom, the glass wafers were first cleaned in Piranha (1:1 H₂SO₄: H₂O₂) for 10 min, followed by RCA1 (5:1:1, DI: NH₄OH: H₂O₂) and RCA2 (5:1:1, DI: HCl: H₂O₂) at 60 °C for 10 min respectively and finally a megasonic cleaning at 60 °C for 10 min. A chromium layer (12 nm) was evaporated as an adhesion layer, followed by a gold layer of 100 nm, deposited in two steps of equal thickness with 10 min pause, without breaking the vacuum inside the evaporator. To define the pattern for connection trenches, a positive photoresist, Microposit S1813 (Dow Chemical Co.) was spun to a thickness of 1 μm and etch marks were etched in the gold layer using a gold etch (100 g KI, 25 g I₂ in 100 g H₂O₂), after which the resist was stripped in acetone.

A positive photoresist, AZ 9260 (MicroChemicals) was sprayed to a thickness of 12 μm and soft baked at 90 °C for 140 s, in which patterns for channels were developed, and the gold was stripped in the openings using the gold etch, followed by a chromium etch [(NH₄)₂Ce(NO₃)₆ (10–15%)+ HCl(15–20%+ H₂O)] prepared by Sunchem AB (Sweden). Before the glass etching, the backside was protected using a surface protection film SWT 10 (Nitto). The inlets and outlets were etched isotropically for 12 min in concentrated HF, resulting in approximately 90 μm deep trenches.

The photoresist and metal masks were stripped on both wafers, and on the channel wafer, a new layer of 12 μm spray photoresist was deposited, same procedure as mentioned above,
in which the channels were patterned. After removing the chromium and gold in the openings, the wafer was etched in concentrated HF for 4.5 min, resulting in a depth of 40 μm, again protecting the backside with the surface protection film. The etch depth was controlled using a Dektak 150 (Veeco) profilometer. After etching, the photoresist and metal masks were stripped.

Prior to bonding, the wafers were cleaned according to the procedure mentioned above, i.e. Piranha, RCA1, RCA2, and megasonic cleaning. To minimize the risk of particles affecting the bond quality, the bonding was done within 10 min after the cleaning. The wafers were manually aligned under a microscope and before the annealing the wafers were pressurized by 1 kg for 60 min. The fusion bonding was done in a vertical furnace at 625 °C. This resulted in temperature variations of approximately 0.1 °C, which caused the flow rate to vary by approximately by 1 μl min⁻¹ at 10 MPa because of the change in density. For the lower flow rates of 10 μl min⁻¹, the thermally induced flow variations of 1 μl min⁻¹ were considered too large. Hence, the compressor was changed to a more powerful RK20 (Lauda), for which the cyclic behavior of the temperature variation was no longer visible and the CO₂ tower was cooled down to 8 °C.

To regulate the chip pressure, a 12.5 m long capillary, i.d. 75 μm, was used for the flow rates of 70 μl min⁻¹, and 1.5 m long capillary, i.d. 25 μm for 10 μl min⁻¹. The chip was attached to a heating stage with resistively heated Kanthal wires using thermal grease, and the temperature was set using a constant power from a power supply QL335TP (TTi).

The temperature was logged at 0.3 Hz at the heating stage and at the top of the CO₂ pump using Pt-100 sensors using a four-wire configuration, and in the H₂O bath at the pressure regulating capillary using a thermocouple type K.

The capillary of the glass chip was connected to the syringe pump tubing using PEEK sleeves and fittings. To prevent air convection from affecting the CO₂ flow, the stainless steel tube supplying the CO₂ from the pump was thermally insulated by a 1 m, i.d. 75 μm capillary, which was not possible to use at higher flow rates because the large flow restriction.

The outlet capillary was submerged in a temperature regulated H₂O bath to provide thermal stability and to be able to change the pressure by varying the bath temperature, i.e. changing the viscosity of the fluid passing through. As an approximation, the pressure drop is determined by the Hagen–Poiseuille equation,

\[
\Delta P = \frac{8\eta Q}{\pi r^4}.
\]

valid for an incompressible fluid. However, using H₂O and CO₂ at high pressures will result in a compressible two-phase flow, which requires a more complex model for describing the pressure drop. Based on the pressure drop from a completely H₂O filled capillary, which decreases approximately 30% for a 1:1 CO₂:H₂O-flow, this approximation was found to be good enough as a rough estimation to determine the outlet capillary length. To prevent ice clogging at the outlet caused by the expanding gas, the capillary end was placed inside a wired heater and set to approximately 90 °C.

The glass chip was imaged using a high-speed camera, Phantom Miro M310 (Vision Research) at 2500 fps.

### 3.3. Procedure

To build up pressure during the initial compression phase of CO₂, a higher volumetric flow rate was used with a higher...
The system pressure, the sample temperature was swept, starting at 50 °C, decreasing in steps of 10 °C down to 25 °C (below the critical temperature). At these conditions, the density for CO2 varied between 350 kg m\(^{-3}\) and 860 kg m\(^{-3}\), and the dynamic viscosity for CO2 varied between 30 μPa s and 83 μPa s. Before each measurement, the logged temperatures and pressures were allowed to stabilize until no trend of increase or decrease in pressure and temperature was seen for at least 10 min.

### 3.4. PLS-modeling

The multivariate analysis was done using Simca-P 13.0 (Umetrics). The general trends in the data set were investigated with PCA. For each set of pressure and temperature at the chip, interpolated values of tabulated data for density [3, 6], viscosity [4, 6], and interfacial tension between CO2 and H2O [38, 39], respectively, and calculated values of Re, We and Ca for both the CO2 flow and H2O flow, were all initially used along with the flow rate of the two pumps, summarized in Table 1. The length of the parallel flow, \(L\), divided by half the

### Table 1. Variables used in PCA model.

<table>
<thead>
<tr>
<th>Variable</th>
<th>Unit</th>
<th>Dependencies</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>(P)</td>
<td>MPa</td>
<td>(P, T_{\text{CO}_2-\text{pump}})</td>
<td>Measured system pressure at the H(_2)O pump.</td>
</tr>
<tr>
<td>(T)</td>
<td>°C</td>
<td></td>
<td>Measured temperature of the glass chip.</td>
</tr>
<tr>
<td>(T_{\text{CO}_2-\text{pump}})</td>
<td>°C</td>
<td></td>
<td>Measured temperature of the CO(_2) pump.</td>
</tr>
<tr>
<td>(\dot{Q}_{\text{CO}_2})</td>
<td>(\mu\text{l min}^{-1})</td>
<td></td>
<td>Measured flow rate of CO(_2) at the pump.</td>
</tr>
<tr>
<td>(\dot{Q}_{\text{H}_2\text{O}})</td>
<td>(\mu\text{l min}^{-1})</td>
<td></td>
<td>Measured flow rate of H(_2)O at the pump.</td>
</tr>
<tr>
<td>(L/w)</td>
<td></td>
<td></td>
<td>Measured parallel length divided by channel width.</td>
</tr>
<tr>
<td>(s/w)</td>
<td></td>
<td></td>
<td>Measured segment size of CO(_2) divided by channel width.</td>
</tr>
</tbody>
</table>

\(\rho_{\text{CO}_2-\text{pump}}\) \(\text{kg m}^{-3}\) \(P, T_{\text{CO}_2-\text{pump}}\) Interpolated density from tabulated data [3] of CO\(_2\) inside the pump.

\(\rho_{\text{CO}_2-\text{chip}}\) \(\text{kg m}^{-3}\) \(P, T\) Interpolated density from tabulated data [3] of CO\(_2\) at the chip.

\(\eta_{\text{CO}_2-\text{chip}}\) \(\text{Pa s}\) \(P, T\) Interpolated viscosity from tabulated data [4] of CO\(_2\) at the chip.

\(\rho_{\text{H}_2\text{O}-\text{pump}}\) \(\text{kg m}^{-3}\) \(P\) Interpolated density from tabulated data [6] of H\(_2\)O inside the pump.

\(\rho_{\text{H}_2\text{O}-\text{chip}}\) \(\text{kg m}^{-3}\) \(P, T\) Interpolated density from tabulated data [6] of H\(_2\)O inside the chip.

\(\eta_{\text{H}_2\text{O}-\text{chip}}\) \(\text{Pa s}\) \(P, T\) Interpolated viscosity from tabulated data [6] of H\(_2\)O at the chip.

\(\gamma\) \(\text{N m}^{-1}\) \(P, T\) Interpolated interfacial tension from tabulated data [32, 33] between CO\(_2\) and H\(_2\)O at the chip.

\(\dot{Q}_{\text{total}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2}, Q_{\text{H}_2\text{O}}\) Calculated total flow rate of the two pumps at the pumps.

\(\dot{Q}_{\text{ratio},\text{pump}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2}, Q_{\text{H}_2\text{O}}\) Calculated ratio of CO\(_2\):H\(_2\)O at the pumps.

\(\dot{Q}_{\text{CO}_2-\text{chip}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2}, Q_{\text{H}_2\text{O}}, P_{\text{CO}_2-\text{pump}}, P_{\text{CO}_2-\text{chip}}\) Calculated flow rate of CO\(_2\) at the chip.

\(\dot{Q}_{\text{H}_2\text{O}-\text{chip}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2}, Q_{\text{H}_2\text{O}}, P_{\text{H}_2\text{O}-\text{pump}}, P_{\text{H}_2\text{O}-\text{chip}}\) Calculated flow rate of H\(_2\)O at the chip.

\(\dot{Q}_{\text{total},\text{chip}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2-\text{chip}}, Q_{\text{H}_2\text{O}-\text{chip}}\) Calculated total flow rate of the two pumps at the chip.

\(\dot{Q}_{\text{ratio},\text{chip}}\) \(\mu\text{l min}^{-1}\) \(Q_{\text{CO}_2-\text{chip}}, Q_{\text{H}_2\text{O}-\text{chip}}\) Calculated ratio of CO\(_2\):H\(_2\)O at the chip.

\(R_{\text{co},\text{CO}_2}\) \(Q_{\text{CO}_2-\text{chip}}, P_{\text{CO}_2-\text{pump}}, P_{\text{CO}_2-\text{chip}}, \eta_{\text{CO}_2-\text{chip}}, \gamma\) Calculated Re for the CO\(_2\) flow.

\(C_{\text{co},\text{CO}_2}\) \(Q_{\text{CO}_2-\text{chip}}, \eta_{\text{CO}_2-\text{chip}}, \gamma\) Calculated Ca for the CO\(_2\) flow.

\(W_{\text{co},\text{CO}_2}\) \(Q_{\text{CO}_2-\text{chip}}, P_{\text{CO}_2-\text{chip}}, \gamma\) Calculated We for the CO\(_2\) flow.

\(R_{\text{co},\text{H}_2\text{O}}\) \(Q_{\text{H}_2\text{O}-\text{chip}}, P_{\text{H}_2\text{O}-\text{pump}}, P_{\text{H}_2\text{O}-\text{chip}}, \eta_{\text{H}_2\text{O}-\text{chip}}\) Calculated Re for the H\(_2\)O flow.

\(C_{\text{co},\text{H}_2\text{O}}\) \(Q_{\text{H}_2\text{O}-\text{chip}}, \eta_{\text{H}_2\text{O}-\text{chip}}, \gamma\) Calculated Ca for the H\(_2\)O flow.

\(W_{\text{co},\text{H}_2\text{O}}\) \(Q_{\text{H}_2\text{O}-\text{chip}}, P_{\text{H}_2\text{O}-\text{chip}}, \gamma\) Calculated We for the H\(_2\)O flow. 

ratio of CO\(_2\) to H\(_2\)O (approximately 2.5:1), until a stable pressure was reached to keep a similar ratio between CO\(_2\) and H\(_2\)O at the outlet capillary, as a H\(_2\)O filled capillary would result in a higher flow resistance, and hence give an initially higher back pressure.

For the 50 measurements performed, two flow rates were investigated, 70 \(\mu\text{l min}^{-1}\) and 10 \(\mu\text{l min}^{-1}\), with the relative ratios between the CO\(_2\) pump and the H\(_2\)O pump of 4:9, 1:1, and 9:4, measured and controlled at the pumps.

Low ratio of CO\(_2\) (4:9) resulted in a rapid breakup of the devices, hence only a limited number of measurements were carried out for this ratio.

For all experiments, the chip pressure is assumed to be equal to the pressure inside the H\(_2\)O pump, because the large inner diameter of the stainless steel tube of 1 mm between the chip and the pump. Being approximately 60 cm long, this will only result in a pressure drop in the order of 0.2 kPa for a pure H\(_2\)O flow of 35 \(\mu\text{l min}^{-1}\). The system pressure varied between 9.5 MPa and 17 MPa by varying the length and temperature of the outlet capillary, and for each set of flow rate and
channel width i.e. the initial flow width, w, figure 1, together with the length of CO2 segments, s, also divided w, were used as responses for making two different PLS models. All variables were normalized and centered, and as part of the modeling process, the variables that influence the responses were found. Some of the investigated variables have covariance, and in the final models, the variables that have the best correlation with the responses were chosen. In total there were 50 observations in the data set. Out of these, 45 observations were randomly chosen for the PLS modeling and the remaining 5 was used for model validation.

4. Results

The dimensionless numbers are presented in table 2 for the CO2 flow and the H2O flow, calculated at the inlets to the main channel in accordance with figure 1. Re is within the non-turbulent regime, i.e. Re << 2000 for both the H2O flow and the CO2 flow at the inlet. For all the dimensionless numbers, the values for the CO2 flow is typically one order of magnitude higher compared to the values for the H2O flow. Three different behaviors were observed; parallel flow through the channel, figure 5(a), parallel flow broken up in segments, figure 5(b), and segmented flow from chip entrance, figure 5(c). For parallel flows extending beyond 1 mm, the CO2-flow was observed to be pushed to the side of the channel at a width estimated at a top down view to be within 30 to 50 µm as seen in figure 5(a).

With PCA, four PCs explain 0.96 of the variation. Together, the two first PCs explain 0.73 and are the most important. The observations in the score plot, figure 6, forms two separate groups depending on flow rate. The flow rate for the large left group was 70 µl min⁻¹ and the smaller right group had a flow rate of 10 µl min⁻¹. The normalized parallel length, L/w, for each observation is shown next to each observation in figure 6, and within the two groups, the observations having longer parallel length have a tendency to have a lower score value t along PC2, i.e. the y-axis. This becomes clearer when the observations is classified into three different categories; e (green) for segments created at the inlet with a parallel length below 0.6 mm, s (blue) for segments created along the channel and p (red) for observations with parallel lengths beyond 9.5 mm. The corresponding loading plot is showed in figure 7, and table 1 lists all the variables in the loading plot. Along PC2, observations with high parallel length L/w is positively correlated to the variables with negative loading values p [2] in figure 6, e.g. Qratio, and ReCO2, while variables like pressure (P) is negatively correlated along same axis. From figure 7, it can also be concluded that a few of the observations where the flow is parallel through the channel overlaps with some points where the segments form half way in the channel. This indicates that the fully parallel regime p is also affected by unknown variables not included in the model. The difference in segment length of CO2, s/w, is instead manifested primarily along PC1, and e.g. a high segment length is positively correlated to a high temperature T and negatively correlated to a high total flow Qtotal.

Table 2. The range of the dimensionless numbers for CO2 and H2O, calculated at the inlet of the channel for the two flow rates.

<table>
<thead>
<tr>
<th>Flow rate (µl min⁻¹)</th>
<th>Re (×10⁵)</th>
<th>Ca (×10⁻⁵)</th>
<th>We (×10⁵)</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO2</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>10</td>
<td>8.5–19</td>
<td>2.9–4.2</td>
<td>29–71</td>
</tr>
<tr>
<td>70</td>
<td>31–147</td>
<td>13–40</td>
<td>640–3800</td>
</tr>
<tr>
<td>H2O</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>10</td>
<td>0.6–1.0</td>
<td>15–27</td>
<td>11–23</td>
</tr>
<tr>
<td>70</td>
<td>3.1–8.8</td>
<td>110–400</td>
<td>530–2150</td>
</tr>
</tbody>
</table>

Figure 6. PCA score plot with the observations divided into three categories; e for segments created at the entrance, s for segments created along the channel, and p for parallel flow throughout the channel. Next to each measurement is the observed parallel length (L/w). The score values (t) along the first PC are plotted along the x-axis t [1], and the score values along the second PC are plotted against the y-axis t [2]. The influence of flow rate is separating the data into two groups along PC1, while the observations arranges according to the categories mentioned along PC2, with a longer parallel length in the negative direction.
Many of the variables co-depend on each other, and during the PLS modelling all variables that did not contribute to a higher quality factor $Q^2$ could be removed without affecting the amount of explained variation.

For the parallel length using 45 of the observations, the best model to predict the normalized parallel length was using $P$, total flow rate at the pumps ($Q_{\text{total}}$), and the ratio between CO2 and H2O at the pumps ($Q_{\text{ratio, pump}}$), for 1 PC according to

$$
\frac{L}{w} = 75.8 - 9.9P + 0.29Q_{\text{total}} + 46.9Q_{\text{ratio, pump}}
$$

The model using these three variables resulted in an explained variation of predicted length ($R^2Y$) of 0.69, and a predictive ability calculated by cross-validation ($Q^2$) of 0.68. Using the five observations in the prediction dataset resulted in a root mean square error of prediction (RMSEP) of 14.6, i.e. a standard deviation of 1.4 mm in the prediction of the normalized parallel length.

Modeling the normalized CO2 segment length ($s/w$) using 34 data points out of the 45 observations where a segment length could be extracted, increasing temperature, decreasing adjusted total flow rate, increasing CO2-ratio and an increasing $We_{\text{H2O}}$ increased the CO2 segment length for 3 PCs according to

$$
\frac{s}{w} = -3.84 + 0.11T - 0.11Q_{\text{total, chip}} + 5.99Q_{\text{ratio, chip}}
 + 414We_{\text{H2O}}.
$$

The model using these four variables resulted in an explained variation of predicted segment length ($R^2Y$) of 0.84, and a predictive ability calculated by cross-validation $Q^2$ of 0.73. Using the 3 prediction observations where the flow actually was segmented resulted in a RMSEP of 0.48, i.e. a standard deviation of 46 μm in the prediction of the normalized segment length.

5. Discussion

The glass surface favors wetting of H2O over CO2, causing instability in the system that eventually leads to H2O wrapping around CO2, and finally pinching off the flow into segments. But the observed parallel lengths were considerably longer compared to what would be the case if this wetting instability limited the parallel length. If this mechanism causes the segmentation, a rough estimation of the length scale before segmentation can be done since Ca roughly describes the velocity of this movement for Ca in the range 0.1 to 1 according to

$$
U \sim 0.1 \frac{Y}{\eta}.
$$

Hence, for the flow width of 95 μm, the pinching movement is expected to occur in the order of 15 μs to 30 μs for the conditions in this study, for which the flows only travels up to
5 μm down the channel. This indicates that other factors affect the break up into segments.

The dimensionless numbers used were calculated at the inlets to the channel. Since the CO2 is pushed to the side, this leads to higher flow velocities for CO2 and the opposite for H2O. The actual Re in the channel will therefore be higher for CO2 and lower for H2O. Attempts to estimate the flow conditions inside the two-phase channel was hampered by only having access to the top down view of the cross sectional area.

In the PCA model, the observations arranged themselves in two groups with respect to the flow rate along the first PC, and further sorted, mostly along the second PC, according to the parallel length. However, for the longest parallel lengths, observation for both fully parallel measurements and segmented flows are overlapping, which indicates an unstable region.

As mentioned in another study [22], high ratios of CO2 at high flow rates are favorable for a parallel flow. Modeling the parallel length using PLS, a longer predicted parallel length was acquired for increasing CO2:H2O ratio, as well as increasing the total flow rate along with a decreasing pressure. Interestingly, an increased predictability of the model was acquired when using the flow rates that were not adjusted for the variations of density. A higher flow rate is also reasonable to associate with a higher parallel length as the inertia is increased for the two fluids, preventing the pinching of the CO2 flow.

For the 10 measurements in the category k, the parallel lengths exceeds the length of the channel for 9 of the measurements which is a drawback when making a model that is able to correctly predict also the longest parallel flows. However, these measurements are important information to have for predicting parallel lengths beyond 40 channel widths, i.e. approximately a third of the channel length. If these measurements are excluded, a model for the shorter parallel lengths can be created, having an explained variation of 0.72 (Q² = 0.7) and for this model, the total flow rate is not longer a significant variable.

In contrast to the model for the parallel length, the segment size is dependent on the temperature at the chip and W_{H2O} as well as the CO2:H2O ratio at the chip together with the flow rate at the chip, where the flow rate is adjusted for the density change from pump to the chip. A higher flow rate will result in longer CO2 segments, as will increasing the ratio of CO2, which is expected. Also, increasing the W_{H2O} increases the segment length, which could be explained by a higher inertia that prevents the pinching movement from breaking up the CO2 flow. The segment length was also possible to predict using W_{CO2}, instead of W_{H2O}, although the use of W_{CO2} resulted in a decrease of approximately 0.04 in explained variance.

Our study does not include the heat transfer of the fluids. Hence, momentarily local variations of temperature during the compression and expansion of the CO2 is not considered.

For our measurement setup, the global temperature is very stable, and the flow from the pumps has very little variation. However, using a passive backpressure regulator that relies on the flow resistance of the fluid passing through, the system becomes less stable as the flow resistance is different for the two fluids. As the actual volumetric ratio of CO2 compared to H2O increases, the flow resistance will decrease reducing the pressure and increasing the volumetric flow rate ratio even further. Hence, it is important to compensate for the different compression of the fluids during the pressure buildup in order to find the pressure giving a constant flow resistance. Working with fluids with different compressibility at high pressures gives this very different behavior, not present working with incompressible fluids. Therefore, the measurements were acquired only after no trend of increase or decrease in pressure was observed for at least 10 min.

The variation of viscosity and density for CO2 becomes smaller at higher temperatures and pressures. As the density increases, so does the solvent power of CO2. For controlled extraction experiments at macro scale, typical pressures and temperatures involving scCO2 is above 15 MPa, and 80 °C, respectively, where the variation in densities are not as large as for the lower regions closer to the critical point. Hence, it would be beneficial to increase the temperatures and pressures. However, the glass chip durability has an upper limit, and in accordance with Tiggelaar [17], attempts to increase the temperature above 55 °C at high pressures have resulted in failures, where the crack almost always occurred at the inlets/outlets. In this study, increasing the pressure above 16 MPa, the devices only survived for a short period of time, seldom enough to stabilize the system, while staying below that pressure the devices could be pressurized for several days.

6. Conclusions

Principal component analysis (PCA) and partial least square regression (PLS) were shown useful to predict and model the behavior of multiphase flows of supercritical carbon dioxide and water.

In the PCA, lower score values along PC2 corresponded to longer parallel length and data that had been classified as parallel, segmented and segmented from the entrance formed groups in the plot.

Using PLS to model the parallel length showed that higher flow rates of CO2 had a positive influence on the parallel length, while a higher pressure resulted in a shorter parallel length.

The length of created CO2 segments were dependent on the temperature together with the flow rate ratio, and volumetric flow rate, both adjusted for the lower density inside the channel. Also, a higher Weber number for the H2O flow resulted in longer CO2 segments.

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