Controlled Inline Fluid Separation Based on Smart Process Tomography Sensors

Benjamin Sahovic1,*, Hanane Atmani2, Muhammad Awais Sattar3, Matheus Martinez Garcia4, Eckhart Schleicher1, Dominique Legendre2, Eric Climent2, Remi Zamansky3, Annaig Pedrono2, Laurent Babout3, Robert Banasiak3, Luis M. Portela4, and Uwe Hampel1,5

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Today's mechanical fluid separators in industry are mostly operated without any control to maintain efficient separation for varying inlet conditions. Controlling inline fluid separators, on the other hand, is challenging since the process is very fast and measurements in the multiphase stream are difficult as conventional sensors typically fail here. With recent improvement of process tomography sensors and increased processing power of smart computers, such sensors can now be potentially used in inline fluid separation. Concepts for tomography-controlled inline fluid separation were developed, comprising electrical tomography and wire-mesh sensors, fast and massive data processing and appropriate process control strategy. Solutions and ideas presented in this paper base on process models derived from theoretical investigation, numerical simulations and analysis of experimental data.

Keywords: CFD simulation, Control systems, Electrical tomography, Inline fluid separation, Wire-mesh sensor

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1 Introduction

Separation processes are present in many industries. In particular, they are vital to the oil industry, e.g., in the petroleum exploration and refinement. In oil exploration, for instance, the separation of the crude oil from other components (water, gas and sand) in the flow coming from the well is achieved by exploiting the difference in the density of these elements. Density-driven separation methods typically employ gravitational or centrifugal effects. Gravity tanks (such as settling tanks or three-phase separators) are devices that rely on gravity to separate phases from a mixture, while cyclones take advantage of the centrifugal force to achieve the separation. Fig. 1 shows simplified schematics of a gravitational separator (a) and a cyclone (b) on the example of oil-water separation.

The vertical velocities of individual rising oil and settling water droplets inside a gravity tank are determined by the balance between the buoyancy, gravity and drag forces. Therefore, this balance determines the time scale of the separation process. Simultaneously, the flow continuously crossing the separator generates a horizontal motion of the particles, defining their residence time inside the device. A successful separation is achieved when the residence time is greater than the separation time. However, this leads to very large devices when small droplets need to be separated as their vertical velocities are low. In the petroleum industry, for instance, gravity separators may be up to 25 m in length and 3 m in diameter [1].

Cyclones are simple and robust alternatives to overcome the size problem of gravitational separators. These devices are designed to generate accelerations much higher than gravity (up to 100g) via a swirling of the mixture. The much higher driving force inside cyclones results in a tremendous decrease in the separation time, which allows a considerable reduction in the size of the equipment. The compactness of cyclones in relation to gravity tanks makes them especially attractive to offshore petroleum exploration, due to the...
spatial limitation of the platforms [2, 3]. Moreover, cyclones are already being used in the petroleum industry, e.g., to remove oil from the water before discharging it into the ocean.

An inline fluid separator is a special type of cyclone separator. The difference is the flow direction of incoming fluids [5–7]. In the cyclonic separator (Fig. 1b), the multiphase flow is entering tangentially while in an inline fluid separator the fluid is entering axially (Fig. 2a). The inline fluid separator has a simpler design and its main parts are the pipe, a swirl element, a pickup tube and control valves.

However, some care must be taken when using cyclones and swirl separators. An oversized swirl motion shatters the oil droplets into smaller elements, which can lead to emulsions. The optimal operation points of swirl flow with respect to separation efficiency are strongly dependent on the individual flow rates (e.g., of oil and water) and on the size of the oil droplets at the inlet [4], which can vary in time. In addition, the swirl flow inside the swirl separator is intrinsically unsteady, leading to a time-dependent and non-optimal separation of the phases.

The unsteadiness of both the inlet flow conditions and the flow dynamics inside the swirl separator makes it desirable to have a real-time control of the device. The complexity of the process and its short time constants are challenges to be overcome. In our concept, the control of the process will be achieved by using tomographic techniques that can provide rich real-time information of the phase distribution upstream and downstream of the swirl element, and a simple reduced-order model of the flow dynamics that can be used for real-time computer control. To test this rather new approach we yet perform experimental designs and analysis on the example of air-water separation, which is somewhat easier to do, but plan to extend the experimental part to oil-water separation later on.

2 Concepts of a Controlled Inline Fluid Separator

Two potential concepts of a controlled inline fluid separator are illustrated in Fig. 2. These concepts are based partly on a fundamental design that was already developed in previous works of Slot [9], van Campen [4], Star [10] and Knöbel [11]. The core element of inline fluid separation is a swirl element inside a pipe that converts pressure into fluid angular momentum via its curved blades. The centrifugal effects, thus, push the heavier fluid towards the wall of the pipe, and the lighter fluid starts to accumulate around the pipe's center line. The lighter phase, which is gas here, is extracted either by a downstream pickup tube (design 1) or via a reverse flow through an extraction channel in the swirl element (design 2). The heavier phase (here water) leaves the separator straight ahead. Upstream and downstream of the swirl element there are tomographic sensors. In the experimental arrangement this will be wire-mesh sensors upstream and an electrical resistance tomography (ERT) sensor downstream the swirl element.

The tomographic sensor upstream of the swirl element is actually a pair of wire-mesh sensors, which allows measuring both gas fraction and gas velocity simultaneously. This provides the control system with information about the incoming flow. The ERT sensor monitors the gas vortex created by the swirl element. The choice of sensors is motivated by the following considerations. The wire-mesh sensor is a very fast instrument with a typical image rate of 10 000 frames per second. It can obtain gas fraction distributions with high rate and accuracy. A further advantage it that it needs less data processing, as it requires, e.g., no image reconstruction. However, the sensor is intrusive and, hence, not suitable for the downstream gas core measurement, as it would create severe interaction with the gas core. There, the ERT sensor as a non-intrusive instrument is much better suited. Moreover, ERT sensors can give some 3D information in principle. However, due to inherent relaxation time constants and image processing, the ERT sensor is somewhat
slower with a rate of about 10 frames per second. This has to be taken into account when designing the control part. More details of the tomographic imaging sensors are given below. The instrumental part is further complemented by pressure transducers upstream and downstream of the swirl element and at the outlets to yield additional dynamic pressure drop information for control. Furthermore, it is foreseen to capture the flow from the gas and liquid outlets and measure the gas carry-under and the liquid carry-over in each stream via scales. A CAD view of the test section is given in Fig. 3 for illustration.

Figure 2. Two conceptual designs of a controlled inline fluid separator with tomographic imaging sensors and a controlled pressure regulator. a) Downstream gas extraction with a pickup tube; b) reverse gas extraction through the swirl element.

Figure 3. Experimental test section for the controlled inline fluid separator.
3 Control Strategy and System

Control of an inline fluid separator is a critical issue as the flow changes are fast and the imaging sensors produce some latency due to image processing and data reduction. Different strategies are being considered for the control step. As a classical controller a model predictive control concept is pursued (Fig. 4). It is based on a combined feedforward (predictive) action based on the wire-mesh sensor signals and a feedback action basing on the ERT sensors operated at a lower frequency. Both the feedforward and the feedback part will be first implemented via black-box transfer functions using sensitivity matrices obtained from experiments. A next step is to convert the black-box into gray-box models using the reduced order modeling described below. Later an extension to other controller types, such as controllers based on deep neural networks is envisaged.

4 Wire-Mesh Sensors

A wire-mesh sensor consists of two planes of parallel wires placed in a pipe cross section [13]. They are separated by a short distance and arranged so that they form an angle of 90° to each other. The phase fraction (ratio of air to water) in the crossing points is obtained via electrical conductivity by successively applying voltages to wires of one plane and measuring electrical current flow to wires of the second plane. The image acquisition rate is 10 000 frames per second and the spatial resolution of the sensors is \( \sim 3 \text{ mm} \).

Fig. 5 shows a photography of the wire-mesh sensors used in this study.

The fundamental parameter received from the wire-mesh sensor is the local gas fraction \( \varepsilon_{ij} \) in each wire crossing. From that, the total cross-sectional gas fraction over time can be computed, which is

\[
\varepsilon(t_n) = \sum_i \sum_j a_{ij} \varepsilon_{ij}(t_n)
\]  

(1)

Here, \( \varepsilon_{ij} \) is the gas fraction in crossing point \((i, j)\), \( a_{ij} \) are weights associated with the crossing points and \( t_n \) is the temporal sampling point. The \( a_{ij} \) mainly account for the round boundary of the wire grid.

Gas-phase velocity can be obtained with a pair of two wire-mesh sensors mounted adjacent to each other with a small axial displacement in the pipe (Fig. 6). To do this, the cross-sectional averaged gas fraction readings from both sensors are cross-correlated giving the cross-correlation sequence

\[
X(k) = \sum_n \varepsilon_1(t_n)\varepsilon_2(t_{n+k})
\]  

(2)

Cross correlation is practically done via fast Fourier transformation (FFT). Its maximum value \( \Delta \varepsilon \) gives the discrete temporal displacement for maximal similarity. Eventually, the average gas-phase velocity \( \bar{u}_g \) is then computed as

\[
\bar{u}_g = \frac{\Delta L}{\Delta k \Delta T}
\]  

(3)

with \( \Delta T = 1/f_s \) being the sampling time step, \( f_s \) being the sampling frequency and \( \Delta L \) being the axial distance between the two sensors. Both gas fraction and gas velocity calculation have been implemented into a fast FPGA hardware for online data processing to provide this information to the control system.

Experiments were performed at Helmholtz-Zentrum Dresden-Rossendorf to study the relationship between upstream and downstream behavior of the two-phase flow in an inline fluid separator using wire-mesh sensor upstream and a camera to observe the gas core downstream. In the investigations it was found that the gas core average diameter is correlated with inlet gas fraction within an uncertainty band of ±15% [14]. This proved that the wire-mesh sensor can be used as a predictor for the forward control.

![Figure 4. Generic control concept for the inline fluid separator.](image-url)

![Figure 5. Wire-mesh sensor.](image-url)
For tomographic imaging in the gas core region, electrical resistance tomography (ERT) is applied. This sensor technology was first introduced for medical applications in 1984 [16]. Recently, this tomography variant has been further developed for process imaging and monitoring [17, 18]. The two most common modalities are electrical capacitance tomography (ECT) and electrical resistance tomography (ERT). As we deal with a conducting liquid, ERT is used.

Here, the main focus is on the detection of the gas vortex and the measurement of its diameter. The tomography principle consists in retrieving cross-sectional information of a media in a noninvasive way [15]. Being noninvasive in this region is very important as an intrusive sensor may disturb the flow and, hence, separation efficiency. A conventional ERT system is composed of three main parts: (i) sensor, (ii) data acquisition electronics and (iii) software for image reconstruction and visualization [19]. ERT sensors are used to obtain conductivity distribution of the flowing mixture and this is achieved by placing sensing electrodes at the periphery of the flow domain, while remaining in contact with the targeted medium [20]. The special excitation measurement scheme, used to obtain pairwise electrical impedances and image reconstruction, is based on a forward model of electrostatic field propagation. A commercialized measurement system Flow Watch together with WebRoc imaging software from Rocsole Ltd. is being used within the inline separation demonstrator. The Flow Watch system can be used for data acquisition and live image reconstruction. The data acquisition rate is 16 Hz and the image reconstruction rate is 4 Hz. For reconstruction, the dynamical Bayesian estimation method is employed. To observe the vortex, it is important to find an appropriate distance from the point of mixing to the point where stable vortex is formed. After some experimental analysis a distance of 500 mm downstream the tip of swirl element was found to be appropriate.

Preliminary experiments with tap water and air in a pipe of 90 mm diameter were conducted at the Liquid-Gas Flow Research Facility of Tom Dyakowski Process Tomography Laboratory (Lodz University of Technology, Poland). The swirl element has been placed 1 m downstream the air and water injection point. Gas pressure and liquid flow rate were varied. Exemplary results for stable gas core are shown in Fig. 8 (pressure between 0.5 and 2.0 bar and liquid flow rates between 10 m³h⁻¹ and 25 m³h⁻¹). As image reconstruction is slow, current work aims at accelerating this by fast linear reconstruction schemes and applying a direct raw data analysis.

Figure 6. a) A pair of wire-mesh sensors used for gas-phase velocity measurement mounted with axial distance ΔL in the flow loop; b) cross-sectional images of largest cross-correlational similarity having a discrete temporal spacing Δk.
6 Fast Data and Image Processing Hardware

Recently, the increasing computational capabilities of computers to process massive data at high rate has made it possible to use tomographic sensors to quantify physical properties of a process in an industrial environment in real time. Tomographic sensors, like wire-mesh sensors and ERT sensors, are generating large data sets. Two main parallel processing architectures are being used for concurrent computation. One of them is the field programmable logic array (FPGA) and the other is the graphic processing unit (GPU). Both differ in how an algorithm is being executed.

The parallel computing platform CUDA (computer unified device architecture) from Nvidia is used to prototype the algorithm for processing of tomographic sensor data. CUDA gives the ability to control a massively parallel processing environment and the algorithms have been tested on a Nvidia RTX 2070 mobile graphic card that has 2304 CUDA cores. Thus, the card can perform 2304 floating-point operations in parallel per cycle.

An FPGA is a semiconductor device that can perform logical operations enabling the user to program its functions. The building block of an FPGA are wires, logic gates and registers. Modern FPGA, such as Arty 7 (Xilinx Inc.), have additional components, like memory blocks, transceivers, protocol controllers, clock generators and even a central processing unit (CPU), which greatly enhance data processing capability.

Eventually, the system will consist of a workstation combining GPU and FPGA capacity. The GPU will be a primary processing engine while the FPGA has a supporting role in the data acquisition. Thus, data from the wire-mesh and tomography sensors will be acquired and processed in real time, so that information on flow properties can be provided in continuous time intervals for control. Therefore, the existing wire-mesh sensor electronics is extended with an FPGA. A subsequent GPU is used to process the sensor data on a high level as well as compute and execute control actions. A similar concept exists for the ERT sensor and feedback control.

7 Process Modeling

7.1 Lumped-Parameter Reduced Order Modeling

Single-phase bounded swirling flows have a case-dependent axial velocity, a negligible radial velocity, and an azimuthal velocity that can be approximated to a solid body rotation close to the center line and a decaying profile in the annular region [23, 24]. The axial, radial and azimuthal velocity profiles can be changed by controlling the flow at the two outlets and/or the geometry of the swirl element. These profiles are used in modeling to track the position of the particles along the flow by assuming a one-way coupling between the phases [4, 25]. The small size of the particles allows the adoption of a tracing behavior (i.e., the velocity of the particle is assumed to be the fluid velocity) in the axial and tangential directions, while considering an additional slip velocity in the radial direction, caused by the centripetal acceleration. The procedure leads to the velocity of a single particle described by:

\[ U_{p, r} = -\left(\frac{4}{3} \left(1 - \frac{\rho_p}{\rho_1}\right) \frac{d_p}{r} \frac{1}{C_d} \right) \left(\frac{1}{r} \right)^{1/3} U_\theta + U_r \] (4)

\[ U_{p, z} = U_z \] (5)

\[ U_{p, \theta} = U_\theta \] (6)

Here, \( U_{p,i} \) represents the components of the particle velocity, \( U_i \) represents the components of the fluid velocity, \( \rho_p \) and \( \rho_1 \) are the densities of the particles and of the liquid phase, \( C_d \) is the drag coefficient and \( d_p \) is the dispersed phase diameter.

Fig. 9 illustrates a particle trajectory along the pipe for two different conditions. The top part shows a scenario where swirl motion is imposed on top of a uniform axial velocity, and no gravity effects are considered. There, the particle has a slip radial velocity in relation to the flow. This condition is modeled in Eq. (4) by the azimuthal velocity of the fluid, that induces a radial slip velocity of the particle.
7.2 Computational Fluid Dynamics

Most facilities or experimental rigs have limited range of operation in which they can perform desired experiments. Often, cost and time prevent excessive experimental studies. To support the modeling of the process on a very detailed level, tools for computational fluid dynamics (CFD) are developed and applied.

In order to simulate the interaction between the two-phase flow and the separator for the description of the gas core formation a hybrid Euler/Lagrange approach is developed. Considering that the scales are ranging from O(1 m), i.e., the length of the device, to O(100–10 μm), i.e., the size of the smallest bubbles, the hybrid Euler/Lagrange method employs an immersed boundary method (IBM) to describe the geometry, a large eddy simulation (LES) for the turbulent flow, a Lagrangian tracking method for the bubble motion and a volume of fluid (VoF) method to describe the gas core region once bubble coalescence takes place after the swirl element. For the numerical simulations the IMFT in-house CFD code JADIM [28] and its solvers for IBM, LES, Lagrangian, and VoF are used. For this study they need to be coupled. This is done via the mass and momentum conservation equations (Navier-Stokes equations) in the following way:

where \( \overline{U}_i \) is the filtered velocity and \( \tau_{ij}^{isp} \) is the subgrid scale stress tensor related to the Smagorinsky coefficient \( C_S \) through a subgrid eddy viscosity.

The dispersed flow is described using the Lagrangian tracking method [32] developed in JADIM. The bubbles are considered as spherical and their motion is described by solving their trajectory equation:

where \( C_M = 0.5 \) is the added mass coefficient, \( V_B \) is the bubble volume and \( U_B \) their velocity. \( F \) is the force exerted by...
the fluid on the bubble and it contains the drag, lift, virtual mass and Tchen forces \cite{32}. These forces are calculated using the fluid velocity interpolated at each bubble position. From $F$ we deduce the resulting force $F_{LAG}$ induced by the bubbles’ motion inside the liquid. An elastic collision is introduced to describe the interaction between the bubbles and the IBM solid walls.

For simulating the gas core, a VoF method is chosen. It tracks the interface through the VoF function $C$, which stands for the volume fraction, i.e., $C = 1$ is pure gas phase and $C = 0$ is pure liquid phase. The motion of the gas-liquid interface is then governed by the transport equation:

$$\frac{\partial C}{\partial t} + U_i \frac{\partial C}{\partial x_i} = 0$$

(11)

The VoF \cite{28} solver in JADIM does not need any interface reconstruction and is based on a flux corrected transport (FCT) solver. The capillarity contribution $F_{\chi}$ in Eq. (7) requires a specific treatment to avoid spurious current \cite{28}.

Let us consider a numerical domain of size $L_x \times L_y \times L_z$ on a Cartesian mesh made of $N_x \times N_y \times N_z$ nodes. The $x$-direction corresponds to the axis of the pipe. The mesh is regular so that the grid size $\Delta$ is uniform along the three directions. The domain considered in the following examples is of size $L_x = 0.92$ m and $L_y = L_z = 0.1$ m. Fig. 10 displays the domain and a simulation result. The top figure shows the separator geometry as used with the IBM, the middle figure shows the streamlines of single-phase flow for Reynolds number $Re = 4600$ and the bottom figure shows the bubbles tracked using the Lagrangian method and IBM.

8 Conclusion

In this paper, the novel idea of an inline fluid separator controlled with the help of a set of tomographic imaging sensors is introduced. For that, different basic designs are feasible and will be assessed and studied in the future. Many aspects have been considered in what would make the separator efficient including separator components, tomographic sensors, control system and data processing units. The control concept shall be based on a combined feedforward control with predictor using a wire-mesh sensor and a feedback control using an ERT sensor. As inline fluid separation is a complex and fast process, it is difficult to build control algorithms on detailed physical modeling. Hence, a combination of process modeling using physical laws and black-box models fitted by experimental data shall be applied.

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Symbols used

- \( a_{ij} \) [-] weight coefficient of crossing points in a WMS
- \( C_S \) [-] local Smagorinsky coefficient
- \( C_d \) [-] drag coefficient
- \( C_M \) [-] mass coefficient
- \( d_p \) [mm] dispersed phase diameter
- \( f_s \) [Hz] sampling frequency
- \( F_{IBM} \) [N] volumetric force from IBM method
- \( F_{LAG} \) [N] volumetric force from Langrangian method
- \( F'_{e} \) [N] volumetric force from VoF method
- \( g \) [m s\(^{-2}\)] earth’s gravitational constant
- \( k \) [-] discrete temporal displacement
- \( \Delta L \) [mm] axial distance between sensors
- \( P \) [Pa] pressure
- \( \Re \) [-] Reynolds number
- \( T_0 \) [-] temporal sampling point
- \( \Delta T \) [s] sampling time
- \( u_{p,i} \) [m s\(^{-1}\)] particle velocity
- \( u_i \) [m s\(^{-1}\)] fluid velocity
- \( v_g \) [m s\(^{-1}\)] average gas-phase velocity
- \( V_i \) [V] voltage of source electrode
- \( V_0 \) [V] ground voltage

Greek letters

- \( \varepsilon \) [-] gas fraction
- \( \rho \) [kg m\(^{-3}\)] density
- \( \nu \) [m s\(^{-1}\)] kinematic viscosity
- \( \sigma^{sgs} \) [-] stress tensor

Sub- and Superscripts

- \( i \) row coordinate
- \( j \) column coordinate
- \( n \) number of samples
- \( g \) gas phase
- \( l \) liquid phase

Abbreviations

- CAD computer-aided design
- CFD computational fluid dynamics
- CPU central processing unit
- CUDA computer unified device architecture
- DSM dynamic Smagorinsky model
- ECT electrical capacitance tomography
- ERT electric resistance tomography
- FCT flux corrected transport
- FFT fast Fourier transformation
- FPGA field programmable gate array
- GPU graphic processing unit
- IBM immersed boundary method
- IMFT Institute of Fluid Dynamics Toulouse
- LES large eddy simulations
- MCX micro coaxial connector
- RF radio frequency
- SGS subgrid scale
- TOMOCON smart tomographic sensors for advanced industrial process control
- VoF volume of fluid
- WMS wire mesh sensor

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**Controlled Inline Fluid Separation Based on Smart Process Tomography Sensors**

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**Research Article:** Multiphase flow entering an inline fluid separation system can be measured fast and precise with a wire-mesh sensor. A vortex-shaped gas core, as a response of such system, can be quantified using an electrical resistive tomography sensor. In this study, both sensors together with a fast data processing hardware are part of a concept to control fluid separation.