Petascale simulations of compressible flows with interfaces\textsuperscript{1}\textsuperscript{2}

F. Wermelinger, U. Rasthofer, P.E. Hadjidoukas, P. Koumoutsakos\textsuperscript{*}

Computational Science and Engineering Laboratory, ETH Zurich, Switzerland

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**A B S T R A C T**

We demonstrate a high throughput software for the efficient simulation of compressible multicomponent flow on high performance computing platforms. The discrete problem is represented on structured three-dimensional grids with non-uniform resolution. Discontinuous flow features are captured using a diffuse interface method. A distinguishing characteristic of the method is the proper treatment of the interface zone as a mixing region of liquid and gas. The governing equations are discretized by a Godunov-type finite volume method with explicit time stepping using a low-storage Runge-Kutta scheme. The presented flow solver Cubism-MPCF is based on our Cubism library which enables a highly optimized framework for the efficient treatment of stencil based problems on multicore architectures. The framework is general and not limited to applications in fluid dynamics. We validate our solver by classical benchmark examples. Furthermore, we examine a highly-resolved shock-induced bubble collapse and a cloud of \( O(10^3) \) collapsing bubbles, which demonstrate the high potential of the proposed framework and solver.

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1. Introduction

Cavitation refers to the rapid growth of vapor cavities in a liquid, followed by a violent collapse due to environmental pressure variations. It is known that cavitation causes material erosion on nearby surfaces, thus considerably reduces the expected lifespan of applications in marine propulsion, turbomachinery or fuel injection engines; see, e.g., [1]. Cavitation bubbles usually appear in a cloud which amplifies the destructive potential compared to the single bubble case, e.g., [2]. Nearby bubbles influence the dynamics of the collapse process compared to a single bubble confinement. Non-spherical collapses with focus towards the center of the cloud evolve, while the cloud collapses from the outer hull towards the center. Energy accumulates due to the earlier collapse of neighboring bubbles which is finally released in the form of a strong pressure wave emerging from the center of the cloud. Experimental results of collapsing bubble clouds were published, e.g., in [3–5]. Such setups pose a serious demand on measurement technologies. Numerical studies of cloud cavitation collapse try to alleviate these difficulties and provide additional insight to support the understanding of results observed in experimental studies; see [6,7].

Our aim is to perform detailed simulations of cavitating bubble clouds for the study of bubble-bubble and bubble-flow interactions. Therefore, we target at clouds with thousands of resolved cavities, that is, two orders of magnitude larger than the ones shown in [7]. Such simulations require not only an accurate model to treat the interface region correctly, but also highly efficient flow solvers to manage the large resolution requirement near interfaces. Diffuse interface methods for compressible multicomponent flow (see, e.g., [8–11]) resolve both fluid components by the computational grid. They do not treat the discontinuities at the interface, such as the jump in the density, in a sharp way, but introduce an artificial zone around the interface where the transition from one component to the other takes place in a smooth fashion. Therefore, these methods enable an adequate compromise between accuracy and computational efficiency. In the preceding study [12], we presented a compressible multicomponent flow solver, named Cubism-MPCF, with smoothed interface that is capable of processing grids with trillions of cells.

In this work, we investigate the performance of diffusive interface methods for compressible multicomponent flow and incorporate them into our open-source software Cubism-MPCF. We elaborate on the beneficial impact on the resolution requirements when properly treating the diffuse interface zone as a mixture of a liquid and a gas and not only as an artificial transition region from one component to the other. This is of paramount importance for achieving the targeted bubble counts. We further extend the work [13] by implementing non-uniform grids in our Cubism-MPCF soft-

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\textsuperscript{2} Corresponding author.

E-mail address: petros@ethz.ch (P. Koumoutsakos).

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ware and widen the discussion about the behavior of the material interface in the discrete domain. To demonstrate the capabilities of the extended solver, we added simulation results of a highly-resolved 3D shock-induced collapse of a single bubble near a solid wall. Results for a cavitation tube test case are added to demonstrate the behavior of the material interface in a flow situation with strong pressure relaxation. Furthermore, we report on simulations of cloud cavitation collapse with $O(10^4)$ cavities.

The paper is structured as follows: The governing equations and their discretization are introduced in Section 2. Section 3 provides an overview of our highly optimized flow solver Cubism-MPCF. Benchmark examples for validation as well as results from a highly-resolved shock-induced bubble collapse and a collapsing cloud of 2500 bubbles are presented in Section 4. We conclude our work in Section 5.

2. Computational method

2.1. Governing equations

The governing equation system, derived from the Baer-Nunziato model [14], reads as

$$\frac{\partial \rho_1}{\partial t} + \nabla \cdot (\rho_1 \mathbf{u}) = 0, \tag{1}$$

$$\frac{\partial \rho_2}{\partial t} + \nabla \cdot (\rho_2 \mathbf{u}) = 0, \tag{2}$$

$$\frac{\partial \mathbf{u}}{\partial t} + \nabla \cdot (\mathbf{u} \otimes \mathbf{u} + \mathbf{p} I) = 0, \tag{3}$$

$$\frac{\partial E}{\partial t} + \nabla \cdot (\mathbf{p} \mathbf{u}) = 0, \tag{4}$$

$$\frac{\partial \mathbf{u}_2}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u}_2 = K \nabla \cdot \mathbf{u}, \tag{5}$$

where

$$K = \frac{\alpha_1 \alpha_2 (\rho_1^2 c_1^2 - \rho_2^2 c_2^2)}{\alpha_1 \rho_1 c_1^2 + \alpha_2 \rho_2 c_2^2}, \tag{6}$$

see e.g., [15, 16] for derivation. This equation system comprises two mass conservation equations, one for each component, conservation equations for momentum and total energy in mixture-(or single-)fluid formulation as well as a transport equation for the volume fraction of one of the two components with source/sink term on the right-hand side, referred to as “K-div term” in the following. In Eqs. (1)–(3), $\mathbf{u}$ denotes the velocity, $p$ the pressure, $I$ the identity tensor, $\rho$ the (mixture) density, $E$ the (mixture) total energy $E = \rho e + \frac{1}{2} \rho |\mathbf{u}|^2$, where $e$ is the (mixture) specific internal energy. Moreover, $\rho_k, \alpha_k$ and $c_k$ with $k \in \{1, 2\}$ are density, volume fraction and speed of sound of the two components. It holds that $\alpha_1 + \alpha_2 = 1$ as well as $\rho = \alpha_1 \rho_1 + \alpha_2 \rho_2$ and $p = \alpha_1 \rho_1 e_1 + \alpha_2 \rho_2 e_2$ for the mixture quantities.

The K-div term, originally derived in [17] for homogeneous mixtures, describes the reduction of the gas-volume fraction in a mixture of gas and liquid when a compression wave travels across the mixing region. For an expansion wave, it recovers the increase of the gas-volume fraction. The K-div term is only non-zero in the interface zone, is frequently neglected (see, e.g., [8, 18, 16]). However, treating the interface zone indeed as mixture of gas and liquid and not only as an artificial blending region from one component to the other is important for properly capturing the dynamics of collapsing bubbles, as already indicated by the results shown in [11]. Furthermore, this term also allows for dynamically creating interfaces, corresponding to the generation of gas pockets in low pressure regions; see, e.g., [10].

The equation system is closed by an appropriate equation of state for each of the pure fluids. To capture liquids and gases, the stiffened equation of state, which enables a simple, analytic approximation to arbitrary fluids, is applied; see, e.g., [19] for a review. It is given by

$$p = (\gamma_k - 1) \rho_k c_k^2 - \gamma_k p_k c_k, \tag{7}$$

where isobaric closure is assumed; see, e.g., [16]. Parameters $\gamma_k$ and $p_k$ depend on the material. For $p_k = 0$ and $\gamma_k$ being the ratio of specific heats, the equation of state for ideal gases is recovered. Unless otherwise specified, $\gamma_1 = 6.59$ and $p_{c,1} = 4.069 \times 10^8$ Pa are used for water and $\gamma_2 = 1.4$ and $p_{c,2} = 0$ Pa for air.

2.2. Numerical method

System (1)–(5) is discretized using the method of lines. Our formulation builds on a finite volume method for uniform and non-uniform structured grids which are constant in time. Spatial operators are approximated using high-order stencil schemes. The approach yields a system of ordinary differential equations

$$\frac{d\mathbf{V}(t)}{dt} = \mathcal{L}(\mathbf{V}(t)), \tag{8}$$

where $\mathbf{V}$ is a vector of cell average values and $\mathcal{L} \cdot$ is a discrete operator that approximates the convective fluxes and the source term in the governing system. The temporal discretization of Eq. (8) is obtained by an explicit third-order low-storage Runge-Kutta scheme proposed in [20]. The computation of the numerical fluxes is based on a Godunov-type scheme using the approximate HLLC Riemann solver introduced for single-phase flow in [21]. The Riemann initial states are determined by a shock capturing fifth-order WENO reconstruction (see [22]). Following [23], the reconstruction is carried out using primitive variables, and the HLLC Riemann solver is adapted to Eq. (5) to prevent oscillations at the interface. As the non-uniform grid remains constant in time, the computation of the varying coefficients for the WENO reconstruction must be carried out once. Relations for the non-uniform WENO coefficients have been presented in [18]. The solution is advanced with a time-step size that satisfies the Courant-Friedrichs-Lewy condition. For the weighting coefficients of the Runge-Kutta stages, the values suggested in [24] are used, resulting in a total variation diminishing scheme.

3. Cubism-MPCF

The computational method presented in the previous section is implemented into Cubism-MPCF. Cubism-MPCF was presented in its original form in [12] and later further developed in [25, 26]. The most important features of Cubism-MPCF are summarized in the following, additional details about design considerations can be found in the aforementioned references.

We define the computational kernels RHS, DT and UP for the processing of the full discretization. RHS stands for “right-hand side” and is responsible for the evaluation of the approximate Riemann problem. DT determines the admissible time-step size based on a global reduction. Finally, UP performs the Runge-Kutta update for a given set of weighting coefficients. The computational cost distribution of RHS, DT and UP is 90%, 2% and 8%, respectively (neglecting I/O operations).

The software is designed for high performance computing (HPC) and achieves a throughput of more than 700 Billion cells per second [12] with a peak performance of 14.4 Plop/s on BlueGene/Q systems [25]. The code relies on our Cubism library which provides a framework for the efficient treatment of high-order stencil schemes (e.g., RHS) as well as straightforward point-wise operations, such as DT and UP. The Cubism framework is general and not limited to
applications in fluid dynamics. It can be used to implement any type of solver on structured Cartesian grids. The library supports uniform and stretched non-uniform grids. The latter is implemented based on a mapping method which supports user-defined functions that describe the non-uniform mesh spacing. The software is open-source and can be downloaded from our lab repository.1

### 3.1. Hierarchical data structure based on static size blocks

Cubism-MPCF makes use of a two-level hierarchical data structure utilizing cubic blocks of static size. Data within the blocks is stored in an array of structures (AoS) format. The computational domain is composed by a set of blocks arranged on a Cartesian topology, maintained by the Cubism library. The static size of the blocks is chosen such that cache utilization is maximized on multicore architectures. The hierarchical structure allows for increased temporal locality of the data, which is a main concern for stencil operations. The vector $\mathbf{V}$ in Eq. (8) makes use of this hierarchical structure. The low-storage Runge-Kutta scheme further doubles the memory footprint for each block. The varying coefficients of the numerical scheme (for a non-uniform mesh) are computed per block and stored in additional 1D arrays for each spatial dimension. This additional memory footprint is about two orders of magnitude smaller compared to the main block data. The pre-computed coefficients significantly reduce the computational overhead of the numerical scheme compared to a more advanced adaptive multi-resolution strategy. Moreover, the overhead due to the additional memory accesses is negligible compared to the uniform case where coefficients can be hard-coded.

### 3.2. Parallelization and optimizations

The solver is parallelized with a hybrid paradigm using the MPI and OpenMP programming models. The software is split into three abstraction layers to separate specific optimizations, increase reusability of the code and allow for more efficient prototyping of new simulation cases. The realization of the domain decomposition and the inter-rank communication is accomplished on the cluster layer. The subdomains are of fixed size and organized on a Cartesian processor topology. The cluster layer is responsible for the dispatch of the grid blocks to the node layer. The performance optimization techniques used for the block processing on the cluster layer are described in [25]. The thread level parallelism (TLP) is exploited on the node layer using the OpenMP standard. We employ dynamic work scheduling and a parallel granularity of one block per thread to hide potential load imbalances during the processing of the dispatched blocks. Each thread exclusively works on one block at a time. The block data and ghosts, required for the evaluation of the stencil at block boundaries, are loaded into a per-thread work buffer. The intra-rank ghosts are obtained from loading fractions of the surrounding blocks, whereas inter-rank ghosts are obtained from a global buffer. The computational kernels are executed on the core layer, which are called from the executing thread on the node. The core layer exploits data level parallelism (DLP) and instruction level parallelism (ILP) by relying on explicit vectorization and code fusion techniques reported in [25,12]. In order to apply these techniques, the core layer transforms the AoS layout into a structure of arrays (SoA) by converting the data into slices arranged on a ring buffer. More details on software design regarding the parallelization strategy used in Cubism-MPCF can be found in [12].

### 3.3. Strong and weak scaling

We evaluate the performance of our solver on the Piz Daint supercomputer at Swiss National Supercomputing Centre (CSCS). The machine consists of 5272 Cray XC30 compute nodes, each equipped with an Intel Xeon E5-2670 8-core SandyBridge CPU and an Nvidia Tesla K20X GPU. The executable code has been generated with the GNU C++ compiler (v4.8). Fig. 1 shows the detailed strong scaling on a single node with hyperthreading enabled. The size of the cubic blocks is set to 32 cells along each edge. Our code exhibits excellent strong scaling up to 8 cores. The additional gain due to hyperthreading is small because of the high per-thread cache utilization. The Craypat profiler reports a 95.3% hit ratio for the L1D cache and 99.5% for the L1D and L2D caches combined. Fig. 1 displays the weak efficiency across 4096 nodes. The work load on each node is composed of 4096 blocks, resulting in a memory footprint of 8 GB per node. The time to solution for this configuration is 18.3 s (one full time step). The code achieves excellent weak efficiency on the (almost) full machine. The loss in efficiency is due to the collective MPI all-reduce operation required to synchronize the admissible time-step size after the DT kernel finishes. The average time required by the block processing algorithm for each stage of the third-order Runge-Kutta scheme is 6.05 s and 6.13 s for 1 and 4096 nodes, respectively. The timings correspond to one evaluation of the RHS kernel on the whole domain. We achieve this almost perfect compute/transfer overlap by employing the methods explained in [25].

### 4. Numerical examples

#### 4.1. One-dimensional problems

##### 4.1.1. Liquid-gas shock tube

A 1D two-component shock tube, as considered, e.g., in [27], is examined first. The initial conditions in the domain $\Omega = [0, 1] m$ read as

$$(\rho, u, p, u_2)_0 = \begin{cases} 
(1000.0 \text{ kg/m}^3, 0 \text{ m/s}, 1.0 \cdot 10^9 \text{ Pa}, 0) 
& \text{for } 0 \leq x \leq 0.75 \text{ m}, \\
(10.0 \text{ kg/m}^3, 0 \text{ m/s}, 1.0 \cdot 10^5 \text{ Pa}, 1) 
& \text{for } 0.75 \text{ m} < x \leq 1.0 \text{ m}.
\end{cases}$$

The parameters of the stiffened-gas equation of state are set to $\gamma_1 = 1.44$ and $p_{c1} = 6.0 \cdot 10^5 \text{ Pa}$ for the liquid in the left part and to $\gamma_2 = 1.4$ and $p_{c2} = 0 \text{ Pa}$ for the gas in the right part. Results are evaluated at time $t = 0.234 \text{ ms}$, comparing them to the curves obtained from an exact two-phase Riemann solver. Fig. 2 exemplarily depicts the solution for the density obtained without the K-div term, as it vanishes for this case due to $\mathbf{V} \cdot \mathbf{u} = 0$ at the interface. All results are in good agreement with the exact solution.

##### 4.1.2. Cavitation tube

A cavitation tube, as e.g., shown in [10,28], constitutes the second example considered in this section and illustrates the action of the K-div term. The initial conditions in the domain $\Omega = [0, 1] m$ are given by

$$(\rho, u, p, u_2)_0 = \begin{cases} 
(1150.0 \text{ kg/m}^3, -100.0 \text{ m/s}, 1.0 \cdot 10^5 \text{ Pa}, 1.0 \cdot 10^{-2}) 
& \text{for } 0 \leq x \leq 0.5 \text{ m}, \\
(1150.0 \text{ kg/m}^3, 100.0 \text{ m/s}, 1.0 \cdot 10^5 \text{ Pa}, 1.0 \cdot 10^{-2}) 
& \text{for } 0.5 \text{ m} < x \leq 1.0 \text{ m}.
\end{cases}$$

The setup represents two rarefaction waves in a liquid that contains a small amount of gas. The parameters of the stiffened-gas

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1. [https://gitlab.ethz.ch/mavt-cse/Cubism](https://gitlab.ethz.ch/mavt-cse/Cubism).
equation of state are set to $\gamma_1 = 2.35$ and $p_{c,1} = 1.0 \cdot 10^6$ Pa for the liquid and to $\gamma_2 = 1.43$ and $p_{c,2} = 0$ Pa for the gas. Results are evaluated at time $t = 1.8$ ms. Fig. 2 displays the result for the gas-volume fraction and the pressure, which compare well to the ones shown in [10]. As the rarefaction waves move in opposite directions, the pressure in the center of the domain drops notably, but remains positive. The decrease of the pressure leads to the expansion of the gas such that a bubble of nearly pure gas is formed in the center. This gas expansion is enabled by the K-div term.

4.2. Single-bubble collapse

An isolated air bubble collapsing in water due to a prescribed jump between the initial pressure in the interior of the bubble, $p_{B,0}$, and the far-field, $p_\infty$, is investigated next. A pressure ratio of $p_\infty/p_{B,0} = 10$ is considered. Initially, a zero-velocity field is assumed. The density of water amounts to $\rho_{1,0} = 1000$ kg/m$^3$ and of air to $\rho_{2,0} = 1$ kg/m$^3$. The initial radius $R_{B,0}$ of the bubble is set to 1 mm. The initial pressure and gas fraction field are given by

\[
\alpha_{2,0} = \frac{1}{2} \left[ 1 - \tanh \left( \frac{r - R_{B,0}}{1.5h} \right) \right] \quad \text{and} \quad p_0 = \begin{cases} p_{B,0} & \text{if } 0 \leq r \leq R_{B,0}, \\ p_\infty + \frac{R_{B,0}}{r}(p_{B,0} - p_\infty) & \text{otherwise}, \end{cases}
\]

where $h$ denotes the cell length and $r$ the distance between a point $x$ of the domain and the center of the bubble; see [11]. The solution of the Keller-Miksis equation [29], which constitutes an ordinary differential equation for $R_B(t)$ in the weakly compressible case, serves as a reference for our investigations. The bubble radius from our 3D simulations is obtained based on the gas-volume fraction $\alpha_2$ via

\[
R_B(t) = \sqrt[3]{\frac{3}{4\pi}} \int \alpha_2 \mathrm{d}\Omega.
\]

Exploiting the symmetry of the problem, the computation domain is a cube of size $\Omega = [0, 15R_{B,0}]^3$ which contains one eighth of the bubble centered at the origin. Symmetry boundary conditions are imposed at the three faces of the cube intersected by the interface, while zeroth-order absorbing boundary conditions are assumed for the remaining faces. A grid refinement study is provided in Fig. 3. The initial bubble radius is resolved by approximately 12 cells for the coarser grid and by 25 cells for the finer one. When using the finer grid, differences between the 3D simulation and the solution of the Keller-Miksis equation are only marginal. Already with the coarser grid, the curve from the 3D simulation is close to the reference data. In addition, results obtained with a formulation neglecting the K-div term, which is also frequently used in literature for compressible multiphase flow (see, e.g., [18,16,12]), are included. A pronounced positive impact of the K-div term is observed for both resolutions. For the rather coarse resolutions used here, the minimum radius is reached at an earlier time and deviates significantly from the solution of the Keller-Miksis equation when the K-div term is omitted. Although convergence towards the reference curve can be stated, significantly higher resolutions would be required if the K-div term is omitted.

Furthermore, Fig. 3 illustrates the influence of the K-div term on the behavior of the interface thickness, which is defined as

\[
d_1 = R_{\theta=0.1} - R_{\theta=0.9}
\]

based on two equivalent bubble radii. These radii are associated with the 0.1- and 0.9-iso-contours of the gas-volume-fraction field. The equivalent bubble radius is defined by $R_{\theta} = h \sqrt[3]{\frac{3}{4\pi}} \sum_{i=1}^{\theta} \lambda_i$.
and uses a shifted phase indicator function $\chi_2$ with threshold value $\theta$, which is given by $\chi_2 = 1$ if $\alpha_2 > \theta$ and $\chi_2 = 0$ otherwise. In the definition of $R_0$, $h$ denotes the cell size and $n_c$ the number of grid cells. The interface thickness increases approximately linearly in time when the K-div term is included. This increase can be almost exclusively attributed to numerical diffusion. Without the K-div term, the interface particularly thickens, reaching its maximum at the time of collapse. These investigations reveal the thickening of the interface when the K-div term is not considered, in turn leading to the time shift in the collapse process. Moreover, neglecting the K-div term appears to “stiffen” the system such that a further compression of the bubble is prevented, and large deviations from the expected minimum radius are observed; see Fig. 3. The stiffened behavior may be traced back to the simple advection of $\alpha_2$. However, to ensure a correct thermodynamic behavior (see [30]), the gas content $\alpha_2$ has to adapt to pressure changes in the mixture. Exactly this behavior is recovered by the K-div term. Therefore, the K-div term has to be considered for the present applications.

4.3. Shock-induced collapse of a single bubble near a solid wall

The shock-induced collapse of a single air bubble near a solid wall is presented next. The problem setup is sketched in Fig. 4. The bubble is initialized with radius $R_{0,0}$ and displaced by a standoff distance $H_0 = 2R_{0,0}$ from the wall. The bubble is surrounded by liquid water, where both fluids are initially at rest with pressure $p_0$. A planar shock wave normal to the $x$-direction is initialized at coordinate $x = 3.15R_{0,0}$ with a strength of $p_1/p_0 = 353$. The liquid speed of sound ahead of the shock is $c_L = 1658 \text{m/s}$ and the pressure is set to $p_0 = 1.0 \cdot 10^5 \text{Pa}$. The initial density of the liquid is set to $\rho_{1,0} = 998 \text{kg/m}^3$ and that of air to $\rho_{2,0} = 1 \text{kg/m}^3$. The material parameter for the stiffened equation of state are set to $\gamma_1 = 1.668$ and $p_{c,1} = 4.103 \cdot 10^8 \text{Pa}$ for the liquid and to $\gamma_2 = 1.4$ and $p_{c,2} = 0 \text{Pa}$ for the air. Both fluids are inviscid and time is expressed in non-dimensional form by $t = t/(R_{0,0}/c_L)$. A similar problem in a 2D domain is discussed in [31]. The extents of the computational domain are given by $[L_x \times L_y \times L_z] = [12 \times 15 \times 30]$, where all spatial dimensions are normalized with $R_{0,0}$. The domain is discretized using 1728 $\times$ 1280 $\times$ 2560 cells with increased resolution in the neighborhood of the bubble, such that the initial bubble radius $R_{0,0}$ is resolved by 250 cells, see Fig. 4. Half of the bubble is simulated using a symmetry boundary condition. The wall is modeled with a reflecting boundary, where the impedance of the wall is assumed infinite. The remaining boundaries are absorbing.

Fig. 5 shows numerical schlieren images in the plane through the bubble center normal to the $y$-coordinate. The four different times illustrate the incident shock ($t' = 1.614$), the impact of the upstream bubble wall on the downstream bubble wall ($t' = 8.289$), the expansion of the generated water–hammer shock ($t' = 8.449$) and the post-collapse phase ($t' = 9.709$). The upstream bubble wall (proximal to the incident shock) starts to move towards the wall upon impact of the incident shock while the downstream bubble wall (distal to the incident shock) continues to remain at rest for a short instant after passing of the incident shock due to the shielding of the bubble. Fig. 6 shows the time history of the bubble wall velocities measured for the point proximal to the incident shock and the point distal to the incident shock. The proximal bubble wall experiences a high acceleration at the impact of the incident shock, followed by a low acceleration regime for about six time units. This slow regime is followed by a fast regime where the proximal wall velocity transitions to supersonic in roughly two time units and impacts on the distal bubble wall at $t' = 8.289$, generating a strong water–hammer shock. The wall velocities are evaluated on the liquid side of the interface for a gas-volume fraction that corresponds to $\alpha_2 = 0.1$. A similar analysis has been presented in [31] where both velocities were reported in the subsonic regime. Our result captures a maximum bubble wall velocity that is in the supersonic regime, which is consistent with experimental results reported for spherical cavities [32].

Fig. 6 shows the time history for pressures, $p_{\text{wall}}$, measured in four points along the $y$-axis on the solid wall. The shock reflection is observed after the time $t' = 3$ where the wall pressure doubles in a point far away from the projected bubble center, while the
Fig. 5. Numerical schlieren images of the incident shock phase (top left) impact of upstream bubble wall on downstream bubble wall (top right), expansion of the water-hammer shock (bottom left) and post-collapse phase (bottom right). The images are computed in the plane through the initial bubble center. Rarefaction waves correspond to continuous gray regions and shock waves correspond to black narrow regions. The interface is shown in red color. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

Fig. 6. Velocity measured in the point on the bubble wall that is closest to the incident shock, denoted by proximal bubble wall, and velocity of the point on the opposite side, denoted by distal bubble wall (left). Pressure measured on the solid wall in four points along the y-axis (right).

Point corresponding to the projected bubble center $y/R_{B,0} = 0$ experiences a delay due to the diffraction of the incident shock upstream. The observed pressure drops are caused by rarefaction waves that are reflected back to the wall from the distal bubble interface. The maximum wall pressure $\max(p_w) = 0.67$ GPa is measured in the center point $y/R_{B,0} = 0$ at time $t^* = 9.342$. The time difference between the water-hammer initiation and the impact on the wall is slightly larger than one time unit since the distal bubble wall moves away from the solid wall during the collapse. The impact of the water-hammer shock on the solid wall causes material erosion and eventually reduces the lifespan of related applications. Moreover, a tensile pressure is measured in the point $y/R_{B,0} = 0$ shortly after the water-hammer impact, which generates secondary cavitation that leads to a further increase of the erosive potential on the wall surface.
4.4. Cloud cavitation collapse

In the following, results from a collapsing cloud comprising 2500 gas bubbles are presented. For the setup of the cloud, bubbles are randomly positioned within a sphere of radius $R_c = 3.5$ mm. The radius of the bubbles is also chosen at random from the interval $[0.10;0.15]$ mm such that the average bubble radius amounts to 0.11 mm. While the positions of the bubbles are obtained from a uniform distribution, a log-normal distribution is used for their radii. The considered cloud contains a gas content of 8.4%, resulting in a cloud interaction parameter of $\beta = 85$ (see, e.g., [33] for a definition). The computational domain $\Omega = [0;10]^2$ mm$^2$ is discretized by 1024$^2$ cubic grid cells. Hence, the bubble radii are resolved by 10 to 15 cells. At the boundaries, non-reflecting, characteristic-based boundary conditions are applied; see, e.g., [34]. The bubble pressure is set to $p_{B,0} = 45$ bar. The ambient pressure is set to $p_\infty = 100$ bar, and the initial pressure field is determined as explained in [7]. Pressurized air with $\rho_{2,0} = 11.352$ kg/m$^3$ is assumed for the gas in the bubble and water with $\rho_{1,0} = 1000$ kg/m$^3$ for the surrounding liquid.

Fig. 7 displays the initial cloud, the evolution of the gas volume $V_2$ and the maximum pressure $p_{\text{max}}$ as well as numerical schlieren images (see [35]) from the center plane of the cloud. Consistent with the experimental findings in [5], the peak of the maximum pressure appears several micro-seconds before the minimum volume is reached. The schlieren images show the formation of a spherical pressure wave, in accordance with the symmetry of the configuration. While the pressure wave propagates through the cloud, it increases in strength such that the highest pressures are observed in the core of the cloud. The pressure wave is accompanied by a subsequent strong deformation of the bubbles. The deformation of the bubbles is caused by inward-pointing micro-jets due to bubble–bubble interactions. As the imposed pressure ratio is rather low, the strength of the individual bubble collapses is relatively weak. Together with the relatively high $\beta$, we encounter a regime of weak collapse strength, but strong bubble interactions. As a result, most of the energy of the pressure jump is transferred into kinetic energy via the micro-jets. Owing to the strong micro-jets, all bubbles get pierced during the collapse process.
5. Conclusions

We have presented a compressible multicomponent flow solver for large scale simulation of cavitation flows. We have used a diffuse interface method to capture the interface on uniform and non-uniform structured grids. The computational approach is based on a Godunov-type finite volume method and has been implemented into our open-source software Cubism-MPCF. The software relies on the Cubism library which provides an HPC framework for the efficient treatment of high-order stencil schemes as well as point-wise operations. We have validated our approach for two classical benchmark examples, a liquid-gas shock tube and a single-bubble collapse. Using the latter test case, we have demonstrated the importance of recovering the compression and expansion of the gas in the interface zone when applying smoothed interfaces to compressible flow problems. Eventually, we have investigated a highly-resolved shock-induced bubble collapse and a cloud of 2500 collapsing bubbles, which have highlighted the high potential of the proposed approach and solver.

Ongoing and future research includes the investigation of collapsing clouds of $O(10^5)$ gas bubbles, a comparative study with respect to approaches based on bubble-particle models, the inclusion of viscous and surface tension effects as well as developments towards software for heterogeneous computer architectures.

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References


Fabian Wermelinger is a PhD student in the Computational Science and Engineering Laboratory at ETH Zurich. He received his MSc in mechanical engineering from ETH Zurich in 2015. His research interests include modeling of multicomponent flows, cavitation phenomena and turbulence. Given the computational complexity of these problems, he is further concerned with software development for high performance computing and data compression algorithms.

Ursula Rasthofer obtained her diploma in aeronautics and aerospace from Technische Universität München (TUM) in 2009, and her PhD in mechanical engineering from TUM in 2015. She has been a postdoctoral researcher at ETH Zurich from 2015–2017. Her research interest focuses on modeling and simulation of turbulent flow, numerical methods for multiphase flow and high performance computing.
Panagiotis Hadjidoukas received his PhD in 2003 from the Computer Engineering and Informatics Department, University of Patras, Greece. Currently, he is a senior researcher and lecturer at ETH Zurich. His research interests include high performance computing on modern architectures and parallel frameworks for numerical optimization and uncertainty quantification. He is recipient of the ACM Gordon Bell Prize for peak performance in 2013.

Petros Koumoutsakos is Professor of Computational Science at ETH Zurich (2000–present). He received his PhD in Aeronautics and Applied Mathematics from the California Institute of Technology (1992). He conducted postdoctoral studies at the Center for Parallel Computing (Caltech, 1992–1994) and at the Center for Turbulence Research (Stanford U./NASA Ames, 1994–1997). Petros has been elected Fellow of the American Society of Mechanical Engineers (ASME), the American Physical Society (APS) and the Society of Industrial and Applied Mathematics (SIAM). He is recipient of the Advanced Investigator Award by the European Research Council and led the team that won the ACM Gordon Bell prize in Supercomputing (2013). In 2016 he was elected Fellow of the Collegium Helveticum.