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Volume-of-fluid simulations in microfluidic T-junction devices: Influence of viscosity ratio on droplet size

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We used volume-of-fluid (VOF) method to perform three-dimensional numerical simulations of droplet formation of Newtonian fluids in microfluidic T-junction devices. To evaluate the performance of the VOF method we examined the regimes of drop formation and determined droplet size as a function of system parameters. Comparison of the simulation results with four sets of experimental data from the literature showed good agreement, validating the VOF method. Motivated by the lack of adequate studies investigating the influence of viscosity ratio ($\lambda$) on the generated droplet size, we mapped the dependence of drop volume on capillary number ($0.001 < \text{Ca} < 0.5$) and viscosity ratio ($0.01 < \lambda < 15$). We find that for all viscosity ratios investigated, the droplet size decreases with increase in the capillary number. However, the reduction in the droplet size with the capillary number is stronger for $\lambda < 1$ than for $\lambda > 1$. In addition, we find that at a given capillary number, the size of droplets does not vary appreciably when $\lambda < 1$, while it increases when $\lambda > 1$. We develop an analytical model for predicting the droplet size that includes a viscosity-dependent breakup time for the dispersed phase. This improved model successfully predicts the effects of the viscosity ratio observed in simulations. Results from this study are useful for the design of lab-on-chip technologies and manufacture of microfluidic emulsions, where there is a need to know how system parameters influence the droplet size. Published by AIP Publishing.

I. INTRODUCTION

Droplet-based microfluidics where fluid volumes down to picoliters are manipulated has witnessed a remarkable growth due to applications in biochemical analysis and material synthesis.1-5 In these applications, it is necessary to produce droplets of controlled size. Droplet size is an extremely important parameter as it controls the efficiency of encapsulation of individual cells and biomolecules6-8 and rates of reaction.9,10 From a fundamental point of view, droplet size dictates the mixing dynamics,1,9-13 flow resistance,14-17 breakup,18-20 coalescence,21-23 and collective behavior.24,25 Since experimental conditions including flow rates, fluid properties, and channel dimensions may vary for different applications,26 it is important to understand the drop formation and develop predictive models of how system parameters influence the droplet size.

A widely used microfluidic geometry for producing droplets is the T-junction device, where the continuous and dispersed phases flow orthogonally meet at a junction producing droplets.27,28 Several parameters have been shown to influence the droplet size in T-junction devices.29-32 These include dimensionless parameters such as flow rate ratio ($Q = \frac{Q_D}{Q_C}$), where $Q_D$ and $Q_C$ are volumetric flow rates of dispersed and continuous phases, respectively), capillary number ($\text{Ca} = \frac{\mu_C U}{\gamma}$, where $\mu_C$ and $U$ are the viscosity and velocity of the continuous phase, and $\gamma$ is the interfacial tension), capillary number of the dispersed phase ($\text{Ca}_D = \frac{\mu_D U_D}{\gamma}$, where $\mu_D$ is the viscosity of the dispersed phase and $U_D$ is the inlet velocity of the dispersed phase), Reynolds number ($\text{Re} = \frac{\rho_C U_C}{\mu_C}$, where $\rho_C$ is the density of the continuous phase and $w_C$ is the width of main channel), viscosity ratio ($\lambda = \frac{\rho_D}{\rho_C}$), and density ratio ($\rho = \frac{\rho_D}{\rho_C}$, where $\rho_D$ is the density of the dispersed phase). In addition, geometrical parameters such as width ratio ($W = \frac{w_D}{w_C}$, where $w_D$ is the width of side channel) and height ratio ($H = \frac{h}{w_C}$, where $h$ is the height of the channel) can also influence the droplet size.

Several experimental studies have investigated the dependence of the drop size on system parameters in T-junction devices.29-36 Among these, here we discuss those that are relevant to our work (see Table I). Garstecki et al. focused on the effect of the flow rate ratio on the droplet size and found a linear relationship between drop length ($L$) and flow rate ratio.30 van Steijn et al. investigated how the dimensions of the main and side channels in the T-junction influence the droplet size.31 Similar to Garstecki et al., for a given T-junction geometry, they also found linear dependence of drop volume ($V$) on the flow rate ratio. However, they showed that at a given flow rate ratio, the drop size also depends on the width ratio ($W$) and height ratio ($H$). In both these studies, experiments were conducted at low capillary numbers ($\text{Ca} < 0.01$). Christopher et al. measured the droplet size across a wider range of capillary numbers ($0.001 < \text{Ca} < 0.5$) while maintaining fixed flow rate ratios.36 In addition to the linear dependency of the droplet size on the flow rate ratio, they found that the droplet size decreases by increasing the capillary number. The parameter range covered by these experimental studies is shown in Table I.

Investigating experimentally, the effect of system parameters on the droplet size has limitations. For example, experimentally it is difficult to produce droplets for viscosity ratios greater than unity, making it unclear how the viscosity ratio

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TABLE I. Experimental studies of droplet formation in microfluidic T-junction devices, whose results on droplet size have been compared to VOF simulations.

<table>
<thead>
<tr>
<th>Study</th>
<th>Ca</th>
<th>Q</th>
<th>λ</th>
<th>W</th>
<th>Focus of study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Garstecki et al., 2006</td>
<td>Ca &lt; 0.01</td>
<td>0.01 &lt; Q &lt; 10</td>
<td>0.01 &lt; λ &lt; 0.1</td>
<td>0.25 &lt; W &lt; 1</td>
<td>Droplet size dependency on Q</td>
</tr>
<tr>
<td>van Steijn et al., 2012</td>
<td>Ca &lt; 0.01</td>
<td>0.1 &lt; Q &lt; 10</td>
<td>0.01 &lt; λ &lt; 0.1</td>
<td>0.33 &lt; W &lt; 3</td>
<td>Droplet size dependency on T-junction geometry</td>
</tr>
<tr>
<td>Christopher et al., 2008</td>
<td>0.001 &lt; Ca &lt; 0.5</td>
<td>0.05 &lt; Q &lt; 4.5</td>
<td>0.003 &lt; λ &lt; 0.167</td>
<td>0.5 &lt; W &lt; 2.5</td>
<td>Droplet size dependency on Ca</td>
</tr>
</tbody>
</table>

TABLE II. Summary of numerical simulation studies on droplet formation in microfluidic T-junction devices.

<table>
<thead>
<tr>
<th>Study</th>
<th>Ca</th>
<th>Q</th>
<th>λ</th>
<th>W</th>
<th>Simulation method</th>
</tr>
</thead>
<tbody>
<tr>
<td>Van der Graaf et al., 2006</td>
<td>0.01 &lt; Ca &lt; 0.08</td>
<td>0.05 &lt; Q &lt; 1</td>
<td>λ = 3.44</td>
<td>W = 1</td>
<td>LBM</td>
</tr>
<tr>
<td>Sang et al., 2009</td>
<td>0.002 &lt; Ca &lt; 0.8</td>
<td>0.05 &lt; Q &lt; 1</td>
<td>0.05 &lt; λ &lt; 0.3</td>
<td>W = 1</td>
<td>VOF</td>
</tr>
<tr>
<td>Kashid et al., 2010</td>
<td>0.0047 &lt; Ca &lt; 0.012</td>
<td>Q = 0.5</td>
<td>0.002 &lt; λ &lt; 0.018</td>
<td>W = 1</td>
<td>VOF, ANSYS fluent</td>
</tr>
<tr>
<td>De Menech et al., 2008</td>
<td>0.001 &lt; Ca &lt; 0.07</td>
<td>0.01 &lt; Q &lt; 2</td>
<td>0.125 &lt; λ &lt; 1</td>
<td>W = 1</td>
<td>Phase field</td>
</tr>
<tr>
<td>Liu and Zhang 2009</td>
<td>0.003 &lt; Ca &lt; 0.06</td>
<td>0.1 &lt; Q &lt; 1</td>
<td>0.0125 &lt; λ &lt; 1</td>
<td>W = 1</td>
<td>LBM</td>
</tr>
<tr>
<td>Sivasamy et al., 2011</td>
<td>0.008 &lt; Ca &lt; 0.025</td>
<td>not reported</td>
<td>λ ∼ 0.04</td>
<td>W = 0.5</td>
<td>VOF, ANSYS fluent</td>
</tr>
<tr>
<td>Yang et al., 2013</td>
<td>0.002 &lt; Ca &lt; 0.056</td>
<td>0.125 &lt; Q &lt; 1</td>
<td>λ = 1</td>
<td>W = 1</td>
<td>LBM</td>
</tr>
<tr>
<td>Hoang et al., 2015</td>
<td>Ca &lt; 0.01</td>
<td>1 &lt; Q &lt; 4</td>
<td>0.01 &lt; λ &lt; 0.1</td>
<td>W = 1</td>
<td>VOF, OpenFOAM</td>
</tr>
</tbody>
</table>

0.001 < Ca < 0.5 | 0.05 < Q < 10 | 0.01 < λ < 1 | 0.33 < W < 3 | VOF, OpenFOAM |

influences the droplet size. This feature is evident from Table I, where studies were limited to a viscosity ratio less than 1. Thus, fixing one control parameter while changing the other parameters is not always guaranteed in experiments, lending access to a narrow system parameter space. Moreover, it is difficult to experimentally measure the three-dimensional flow fields and fluid stresses during drop production in T-junction, although some progress has been made.37–40

Numerical simulations provide a unique opportunity to complement experimental investigations. As shown in Table II, several numerical simulations have been pursued to investigate droplet production in T-junction geometries.37–45 These multiphase simulation approaches include volume-of-fluid (VOF),46–48 phase field,49–52 and lattice-Boltzmann methods (LBM).53–56 Most of the numerical simulations to date have been focused on investigating the mechanism of drop formation at T-junction.37–40,42,43 These studies showed that as the dispersed phase enters and fills the main channel, the upstream pressure increases due to blockage of the channel. This built-up pressure along with shear force exerted on the dispersed phase causes the break-up of the dispersed phase. Furthermore, it was found that at higher capillary numbers, the contribution of shear force to drop formation increases, while built-up pressure decreases.

Despite several experimental studies and numerical simulations investigating droplet production in T-junction devices, additional studies are warranted due to the following reasons. First, even though simulation efforts have provided useful insights into the mechanism of drop formation in T-junction geometries, much less attention has been devoted to comprehensive investigations of the influence of system parameters on the droplet size. Specifically, simulations targeting the direct comparison of different sets of experimental data on a droplet size have not been pursued to date. Second, the role of fluid viscosities on a droplet size has not been fully elucidated. In both experiments and simulations (cf. Table II), so far studies have been limited to a viscosity ratio less than unity. Given that for λ > 1, the viscous stress of the dispersed phase can impact drop production; studies covering a broader range of viscosity ratios need to be pursued.

In this study, we use numerical simulations based on the VOF method to study the influence of system parameters on a droplet size (see Table II). In the first part of our study, we carried out numerical simulations to predict regimes of drop formation and the generated droplet size. We compared the simulation results with data from experimental studies shown in Table I, allowing us to assess the capability of the VOF method and validate it. In the second part of the study, we investigated the influence of viscosity ratio on a droplet size and found it to play an important role. Subsequently, we develop a model that predicts the droplet size considering the effect of viscosity ratio.

II. NUMERICAL SIMULATION

A. Volume-of-fluid method and its implementation

A three-dimensional simulation of droplet formation in T-junction geometries was performed using the volume-of-fluid (VOF) method. VOF is an Eulerian method of multiphase flow simulations where fluid properties such as viscosity and density are smoothed and the surface tension force is distributed over a thin layer near the interface as a body force. In VOF, a phase fraction parameter, α, is used to indicate the presence of each phase at every location of the simulation domain. In our simulation, α = 1 for phase 1 (i.e., continuous phase), α = 0 for phase 2 (i.e., dispersed phase), and 0 < α < 1 in the interface region. In VOF, the governing equations including continuity (Eq. (1)), momentum balance (Eq. (2)), and phase fraction equations (Eq. (3)) are solved.
The velocity of each phase has a contribution to the convection of the fluid. In Eqs. (1)–(3), \( U \) is the velocity vector field, \( p \) is the pressure field, \( T \) is the deviatoric stress tensor \( (T = 2\mu S - 2\mu(\nabla \cdot U)I/3) \), \( S \) is the symmetric deformation tensor, \( I \) is identity matrix, and \( f \) is gravitational force. Parameters \( \mu_b \) and \( \rho_b \) are bulk viscosity and density which are based on the weighted average of the distribution of the phase fraction,

\[
\mu_b = \alpha \mu_C + (1 - \alpha) \mu_D, \\
\rho_b = \alpha \rho_C + (1 - \alpha) \rho_D.
\]

The last term on the right-hand side of Eq. (2), \( \mathbf{F}_S \), represents the continuum surface tension force (CSF)\(^{57,58} \) and is nonzero only on the interface. This force term is defined as \( \mathbf{F}_s = \gamma \kappa (\nabla \alpha) \) where \( \kappa \) is curvature \( (\kappa = \nabla \cdot (\nabla \alpha/|\nabla \alpha|)) \).

We solved the momentum balance equation in conjunction with the continuity equation using the Pressure Implicit with Splitting of Operator (PISO) method.\(^{58} \) In this method, the velocity field is predicted and then corrected to advance the pressure and velocity fields in time. In this work we used three PISO iterations. Once the velocity field was found, Eq. (3) was solved to find the phase fraction. Even though Eqs. (1)–(3) yield velocity and phase fraction at every cell in the domain, the location of the interface needs to be identified with high resolution. To achieve this, we used a two fluid formulation where the contribution of each phase to the velocity of the interface is considered, i.e.,

\[
\frac{\partial \alpha}{\partial t} + \nabla \cdot (\alpha \mathbf{U}_C) = 0,
\]

\[
\frac{\partial (1 - \alpha)}{\partial t} + \nabla \cdot ((1 - \alpha) \mathbf{U}_D) = 0,
\]

where \( \mathbf{U}_C \) and \( \mathbf{U}_D \) are velocity vector fields of continuous and dispersed phases, respectively. Here, we assumed that the velocity of each phase has a contribution to the convection of interface based on their phase fraction, i.e.,

\[
\mathbf{U} = \alpha \mathbf{U}_C + (1 - \alpha) \mathbf{U}_D.
\]

Equation (6) can be rearranged and used as the phase fraction equation

\[
\frac{\partial \alpha}{\partial t} + \nabla \cdot (\alpha \mathbf{U}) - \nabla \cdot ((1 - \alpha) \mathbf{U}_r) = 0,
\]

where \( \mathbf{U}_r = \mathbf{U}_D - \mathbf{U}_C \) is called as the compression velocity. Eq. (9) has a new convective term, compared to Eq. (3). This term is only present at the interface and vanishes in the pure phases. The compression velocity is given by

\[
\mathbf{U}_r = \mathbf{n}_f \min \left[ C_\alpha \frac{\varphi}{S_f}, \max \left( \frac{|\varphi|}{|S_f|} \right) \right],
\]

where \( \mathbf{n}_f \) is the cell normal flux and the ratio \( \frac{|\varphi|}{|S_f|} \) is the magnitude of velocity where \( \varphi \) and \( S_f \) are the cell face volume flux and surface area, respectively. \( C_\alpha \) is a user-specified compression factor that can vary from zero to four. Using larger compression factor results in a thinner interface. However, Hoang et al. who studied microfluidic droplet break-up showed that compression factor greater than one causes parasitic currents at the interface,\(^{44} \) therefore, we chose \( C_\alpha = 1 \) in this study.

In order to assure stability and convergence of the simulation we used an adaptive time step method. At the beginning of each iteration, a new time step was calculated based on the Courant number (Co), which is defined as

\[
C_{o} = \frac{|\mathbf{U}_f \cdot S_f|}{d_f \cdot |\mathbf{S}_f| \Delta t},
\]

where \( d_f \) is the distance between two neighbor cells, \( \mathbf{U}_f \) is the velocity at the surface of the cell, and \( \Delta t \) is time step. Then based on the identified courant number, a new time step size was calculated in order to keep Co less than a predefined limit. The value of the Courant number indicates how much each fluid element is displaced over one time step. For example, when \( Co = 1 \) it means that each element of the fluid moves in a distance of one grid size in one time step. The Courant number was evaluated at each computational cell including bulk fluid and the interface in each iteration.

All the numerical simulations were performed in the open-source code OpenFOAM. VOF was implemented in the interFoam solver as reported by Deshpande et al.\(^{60} \) Several works have used interFoam for solving incompressible two phase flows.\(^{19,44,61-63} \) Among those papers Hoang et al. and Nieves-Ramach et al. have used this solver for the simulation of two phase flows at micro-scale.\(^{19,44,61} \) In OpenFOAM, governing equations were discretized by the finite volume center-based method. Boundedness of the phase fraction parameter was controlled by a Total Variation Diminishing (TVD) method implemented in OpenFOAM called MULES (Multidimensional Universal Limiter with Explicit Solution).\(^{44} \)

### B. Simulation setup

The schematic of T-junction’s geometry used in the simulation study is shown in Figure 1. The system has two inlets for continuous and dispersed phases and one outlet. We set \( \alpha = 0 \) at the inlet of the dispersed phase and \( \alpha = 1 \) at the inlet of the continuous phase. A constant velocity was set for both the continuous phase. A constant velocity was set for both phases at the inlet of the system. No slip boundary condition was applied at the walls. All the simulations were carried out for a density ratio, \( \rho = 1 \) and the Reynolds number is \( Re < 0.1 ; \) therefore, inertia is negligible. In this study, time is scaled by \( \frac{\nu}{U_0} \).

In order to predict the values of the phase fraction near the walls, i.e., the interface, the normal of the interface is related to the wall’s normal and tangential unit vectors,

\[
n_{\text{interface}} = \cos(\theta) n_{\text{wall}} + \sin(\theta) t_{\text{wall}},
\]

where \( n_{\text{interface}} \) is the normal of the interface, \( \theta \) is the wall’s static contact angle, and \( n_{\text{wall}} \) and \( t_{\text{wall}} \) are the unit normal and unit tangential vectors to the wall, respectively. The walls of the channels were considered as fully wetted by the continuous
phase; therefore, we set the contact angle as zero at the walls in this study. The boundary condition at the outlet was set as atmospheric pressure.

The simulation domain was meshed using hexahedral cells. Adjacent cells to the wall were refined five times to capture the lubrication film around the droplet. The domain meshing was carried out using ANSYS Gambit version 2.4.6. Since determining the size of the droplets is the main goal of this study, the grid size of the mesh becomes important. We carried out numerical simulations on a representative T-junction geometry to evaluate the sensitivity of the grid size $d$ on the droplet volume. The width and height ratios were set to be $W = 0.5$ and $H = 0.33$, respectively. We performed simulations for a fixed flow rate ratio ($Q = 0.5$) and two capillary numbers ($Ca = 0.01$ and $0.07$). We measured the droplet volume $(V)$ by counting the number of grid points in which $\alpha \leq 0.5$. The results are shown in Fig. 2. We find that for $w_C/d > 50$, the variation in the droplet size is smaller than 4%, while for $w_C/d > 75$ it is smaller than 1%. Therefore, a grid size of $w_C/d = 100$ was chosen in this study.

To clarify whether the upstream and downstream channel lengths of T-junction are long enough for the flow to become fully developed, we computed the entrance length in our system. The entrance length for channel flows under laminar conditions ($Re < 2000$) is given by

$$L_e = d_h \left( \frac{0.6}{1 + 0.035 Re} + 0.056 Re \right),$$

where $d_h$ is the hydraulic diameter of the channel. We find that the maximum entrance length, corresponding to highest $Re$, needed to achieve a fully developed flow is just a small fraction of the upstream ($L_e = 0.11 L_{\text{Upstream}}$) and downstream lengths ($L_e = 0.02 L_{\text{Downstream}}$) used in the simulation. Therefore, the choice of entrance lengths used in the simulation is sufficient to obtain a fully developed flow.

All the simulations were performed in parallel on a Linux cluster (Hrothgar cluster at High Performance Computing Center, Texas Tech University and Stampede cluster, Texas Advanced Computing Center) by employing 48 to 96 processors. Based on initial test simulations, we determined that in order to prevent spurious currents and nonphysical behavior of interface, the courant number and interface courant number need to be kept below 0.1. Typical clock time for the formation of one droplet ranges from 22 to 40 h depending on the number of processors used.

III. RESULTS AND DISCUSSION

A. Regimes of drop formation

Previous studies have shown that three distinct regimes—squeezing, dripping, and jetting—exist for the dispersed phase behavior in a T-junction geometry. In this section, we perform simulations to examine whether VOF can capture the different regimes of drop production. We also compare our simulation results to the experimental data of Tice et al. who identified these regimes as a function of flow rate ratio and capillary number.

Similar to experiments, we observe the three behaviors in our VOF simulations as shown in Fig. 1. We find that in the squeezing regime, the dispersed phase blocks the main channel significantly and breakup occurs in the vicinity of the T-junction. In the dripping regime, the dispersed phase only blocks the main channel partially and penetrates further from the T-junction and the droplets are produced at a fixed...
spatial location in the microchannel. In contrast to the dripping regime, in the jetting regime, we find that after the generation of the first droplet the thread of the dispersed phase continues to move forward and a second droplet is produced further downstream. At high capillary numbers, \( Ca > 0.1 \), we observed that dispersed phase co-flows with the continuous phase forming a parallel stream without any droplet being formed. These different behaviors we observed with VOF simulations are consistent with those observed in experiments,\(^{11,27,67}\) phase field,\(^{38}\) and lattice-Boltzmann simulations.\(^{68}\)

To test whether our VOF simulation can quantitatively predict the transitions between these regimes as a function of system parameters, we compared the results from our simulation with the experimental data of Tice et al.\(^{11}\) In their study, experiments were conducted for a geometry with \( W = H = 1 \), for two different viscosity ratios \( \lambda = 0.1 \) and 10, and for \( 0.1 < Q < 10 \) and \( 10^{-3} < Ca < 10^{-1} \). Simulations were performed at the same parameter values as the experiments.

Figure 3 shows the map of the regimes as a function of the two control parameters—\( Ca \) and \( Q \). A good agreement is observed between the simulation results and the experimental data. As shown in Figure 3, for a relatively small capillary number and flow rate ratio, droplets are produced in the squeezing regime. However, by increasing \( Ca \) and/or \( Q \), the regime of drop formation changes to dripping and jetting. The dripping region is much narrower compared to squeezing and jetting. By increasing the flow rate ratio, we find that the transition from squeezing to dripping and jetting occurs at a lower capillary number. This observation indicates that at higher flow rate ratios, dispersed phase penetrates much more into the main channel such that the continuous phase can break it up only in the downstream region.

By comparing results for two different viscosity ratios in Fig. 3, we observe that the boundaries defining the transition between regimes are steeper for \( \lambda = 10 \) than \( \lambda = 0.1 \). This means that the transition between regimes occurs at significantly smaller capillary numbers for a higher viscosity ratio. When the viscous force of the dispersed phase increases, it is difficult for the continuous phase to fragment it. The viscous force of the dispersed phase depends on both the viscosity and velocity of the dispersed phase. Therefore, for high viscosity ratios or high velocities of the dispersed phase, drops are generated in the jetting regime.

**B. Validating VOF simulation using experimental data of droplet size**

In Sec. III A, we have quantified the regimes of drop formation in the T-junction device with VOF simulations. We note that it is difficult to define sharply the transition between the regimes. Therefore, in order to benchmark the capabilities of VOF simulations more precisely, we compared VOF predictions of a drop size against experimental data. As mentioned earlier, we choose the three experimental data sets reported in Table II to validate the VOF method.

Figure 4 compares the data reported by Garstecki et al.\(^{19,30}\) and the numerical simulations in a fixed T-junction geometry at two different viscosity ratios, \( \lambda = 0.01 \) and 0.1. The experiments were conducted at three different continuous phase flow rates, \( Q_c = 0.0028, 0.028, \) and 0.28 \( \mu l/s \). Garstecki et al.\(^{30}\) plotted the droplet length normalized with the main channel width, \( L/w_c \), as a function of the flow rate ratio, \( Q \). They found that the droplet length is mostly constant at low flow rate ratios but increases almost linearly at high flow rate ratios. Our simulation results also show similar trends and are in good agreement with their experimental data.

The experimental data obtained by Garstecki et al.\(^{19,30}\) in Fig. 4 pertain to a single T-junction geometry. To explore the capability of VOF simulations to predict the droplet size generated in other T-junction geometries, we relied on the experimental study by van Steijn et al.\(^{31}\) They used three different geometries with different width and height ratios, \((W, H) = (0.33, 0.33), (0.67, 0.17), \) and \((1.33, 0.11)\). We used the same geometries in our study and the same viscosity ratio of \( \lambda = 0.1 \). We varied the flow rate ratio \((0.2 < \lambda < 6)\) by fixing the continuous phase flow rate and changing the flow rate of the dispersed phase, van Steijn et al. plotted normalized droplet volume, \( V/w_c^2 \), as a function of the flow rate ratio for the three mentioned geometries (see Fig. 5). They found that the width of the side channel influences the volume of droplet significantly. At a fixed flow rate ratio, the volume of produced droplet in a T-junction with a wider side channel is larger. For example, at \( Q = 5 \) the droplet volume in T-junction with \( W = 1.33 \) is four

**FIG. 3.** Regimes of drop formation in a microfluidic T-junction device with \( W = 1, H = 1 \). VOF simulations (closed symbols) were conducted for (a) \( \lambda = 0.1 \) and (b) \( \lambda = 10 \). Experimental data (open symbols) are from Tice et al.\(^{11}\) The dashed lines are drawn to guide the eye.

**FIG. 4.** Effect of flow rate ratio on the size of droplets for (a) \( \lambda = 0.01 \) and (b) \( 0.1 \). Experimental data (open symbols) are from Ref. 30 and the VOF data are represented by closed symbols. The geometrical parameters are \( W = 0.5, H = 0.33 \) and the flow rates of continuous phase are \( 0.0028, \) 0.028, and 0.28 \( \mu l/S \). The prediction based on Eq. (13) (see main text) is shown by (−).
times larger than in \( W = 0.33 \). As shown in Fig. 5, our simulation results predict the same trend of data as the experimental findings.

The experimental data of Garstecki et al. and van Stein et al. pertain mostly to low capillary numbers (\( \text{Ca} \approx O(10^{-3}) \)). We therefore conducted VOF simulations at high capillary numbers and compared the results with that of the experimental data reported by Christopher et al.\textsuperscript{29} They employed one set of fluids, \( \lambda = 0.01 \) and one geometry, \( W = 1 \) and \( H = 0.33 \) in their experimental study. They fixed the flow rate ratios at \( Q = 0.05, 0.25, \) and \( 0.5 \) and changed the capillary number by varying the inlet velocity of the continuous phase. We performed numerical simulations at the same parameter values as experiments. As shown in Figure 6, our numerical findings are consistent with the experimental data. By increasing the capillary number for any fixed value of flow rate ratio, the droplet volume decreases. On the other hand, by increasing the flow rate ratio at a fixed capillary number, the volume of droplet increases.

Overall, comparison of simulation results with the three sets of experimental data reveals that our VOF methodology and choice of simulation parameters (e.g., grid size, compression factor, and Courant number) are suitable for making measurements of a droplet size in microfluidic T-junction devices.

### C. Importance of fluid viscosity ratio on droplet size

In this section, we discuss a popular model for predicting the droplet size at a low capillary number in T-junction devices, as originally proposed by Garstecki et al.\textsuperscript{30} Given that this simple model ignores the effect of the fluid viscosity ratio, we subsequently study the influence of the viscosity ratio using VOF simulations and highlight the need to incorporate this parameter into analytical predictions of the droplet size.

At low capillary numbers, Garstecki et al. assumed that a drop is produced in a two-stage process. In the first stage, the dispersed phase starts to enter the main channel and occupies a portion of the channel. This stage is called the filling stage. In the second stage, called the squeezing stage, the neck which connects the drop to the dispersed phase is squeezed by the continuous phase, ultimately pinching-off the dispersed phase. Figure 7 shows snapshots from the VOF simulation depicting these two stages of droplet generation.

Using this conceptual picture, Garstecki et al. proposed scaling arguments to obtain a prediction for the droplet size.\textsuperscript{30} They argued that during the filling stage, the extent to which the dispersed phase fills the main channel is approximately equal to the width of the main channel, \( w_C \). In the squeezing step, the droplet is being filled by an amount equal to \( U_D t_{sq} \) where \( U_D \) is the dispersed phase velocity and \( t_{sq} \) is the time needed to squeeze and rupture the dispersed phase finger. By adding these two contributions to the overall droplet length, we have \( L = w_C + U_D t_{sq} \). Assuming that the dispersed phase is being squeezed by a rate proportional to the continuous phase inlet velocity (\( U_C \)) gives an estimate of \( t_{sq} = \frac{d_C}{U_C} \), where \( d_C \) is an undetermined length scale. Using this estimate of \( t_{sq} \), the equation for the droplet length can be rewritten as

\[
\frac{L}{w_C} = 1 + \beta Q, \quad (14)
\]

where \( \beta = d_C/w_C \approx O(1) \) is a fitting parameter that depends on the T-junction geometry. For example, for the data in Fig. 4, \( \beta = 1.13 \pm 0.16 \) and \( 1.67 \pm 0.41 \) for \( \lambda = 0.01 \) and \( 0.1 \), respectively. The effect of the T-junction geometry was explicitly considered by van Steijn et al., as reflected by the data of Fig. 5.
The low capillary number model of Garstecki et al. relates the droplet size to simply the flow rate ratio and does not consider the fluid viscosity ratio. Likewise, the compilation of studies in Tables I and II did not cover a broad range of viscosity ratios and was conducted with fluids for which $\lambda \leq 1$. We therefore investigated the importance of the viscosity ratio by varying $\lambda$ from 0.01 to 15. In the simulations, $\lambda$ was varied by changing the viscosity of the dispersed phase and keeping the continuous phase viscosity constant. Additionally, simulations were conducted at a fixed flow rate ratio ($Q = 0.3$) in a device geometry with $W = 1$ and $H = 0.33$. We chose this flow rate ratio as it allows a wider access into the squeezing and dripping regimes. However, at higher viscosity ratios ($\lambda = 10, 15$), we obtained limited data because the operating window for the squeezing/dripping regime is small as discussed earlier in Section III A. The capillary number corresponding to these simulations was $0.001 < Ca < 0.02$.

Fig. 8 shows the droplet size as a function of $Ca$ for different viscosity ratios. For all viscosity ratios, by increasing the $Ca$, the size of droplet decreases suggesting that increasing the viscous stress of the continuous phase produces smaller droplets. However, the reduction in the droplet size with $Ca$ is stronger for $\lambda < 1$ than for $\lambda > 1$. In addition, we find that at a given $Ca$, the size of droplets does not vary appreciably when $\lambda < 1$, while it increases when $\lambda > 1$.

To understand the observations on the effect of the viscosity ratio, we followed the time evolution of the width of the dispersed phase neck ($d_{\text{neck}}$) during the breakup process, as shown in Fig. 9(a). Here $d_{\text{neck}}$ is measured at the channel location where ultimately the pinch-off occurs. Fig. 9(b) shows how the non-dimensional neck thickness, $d_{\text{neck}}/w_D$, varies with time for $\lambda = 0.1$ and 5. As expected, we observe that the neck thickness increases and then decreases with a maximum. We define the duration in which the neck thickness continues to increase and then decrease as $t_{\text{fill}}$ and $t_{\text{sq}}$, respectively, since these two time scales effectively correspond to the filling and squeezing stages in the droplet generation process.

From the data in Fig. 9(b), we observe that at a fixed capillary number, the duration of filling stage for the two viscosity ratios is the same while the duration of the squeezing stage is different. In fact, as shown in Fig. 9(c), when explored across the entire viscosity range ($0.01 \leq \lambda \leq 15$) for varying capillary number, we find that $t_{\text{fill}}$ is independent of the viscosity ratio and is larger when the capillary number is low. In contrast, $t_{\text{sq}}$ increases with viscosity ratio but decreases with increase in capillary number. The observed dependence of $t_{\text{fill}}$ and $t_{\text{sq}}$ on $Ca$ and $\lambda$ help to explain our simulations results shown in Fig. 8. Since both $t_{\text{fill}}$ and $t_{\text{sq}}$ decrease with $Ca$, the drop volume decreases with increase in $Ca$. At a fixed $Ca$, since $t_{\text{sq}}$ increases with viscosity ratio, while $t_{\text{fill}}$ is constant, we observe a larger droplet size for a higher viscosity ratio.
D. Improved model for predicting droplet size

Our results thus far highlight that the viscosity ratio affects the dispersed phase breakup behavior and that the squeezing duration increases with the viscosity ratio, leading to larger droplet sizes. The drop breakup model of Garstecki et al. discussed earlier (see Sec. III C) estimates the squeezing time as \( t_{sq} = \frac{d}{d_C} \). Clearly, the effect of the viscosity ratio is missing in this estimate of squeezing time. Here we address this gap and develop an improved model for predicting the generated droplet size.

In our model, similar to previous work,\(^{30,31}\) we assume that the overall drop volume is resulting from volumetric contributions during the filling and squeezing stages, i.e.,

\[
V = V_{\text{fill}} + V_{sq}. \tag{15}
\]

To estimate the volume of the droplet at the end of the filling period, we use Eq. (15), which was derived by Christopher et al.\(^{29}\) Eq. (15) was obtained by conducting a force balance that includes continuous phase shear stress, upstream fluid pressure, and Laplace pressure jumps across the front and rear of the droplet,

\[
\left(1 - \frac{V_{\text{fill}}}{hw_d^2} \right)^3 = Ca \sqrt{\frac{V_{\text{fill}}}{hw_d^2}} \tag{16}
\]

Eq. (15) is consistent with our simulation result that \( t_{\text{fill}} \) is independent of the viscosity ratio but dependent on the capillary number (cf. Fig. 9). In addition, as \( Ca \to 0 \), Eq. (15) predicts that the non-dimensional fill volume \( \frac{V_{sq}}{hw_d^2} \to 1 \), which is consistent with the low capillary number model of Garstecki et al.

To estimate \( V_{sq} \), we determine the overall squeezing time (\( T_{sq} \)) by summing the break-up time of a viscous thread (\( t_b \)) and the squeezing time considered by Garstecki et al., i.e.,

\[
T_{sq} = t_{sq} + t_b. \tag{17}
\]

To determine \( t_b \), we consider the break-up dynamics of Newtonian filaments that has been investigated in several studies.\(^{69–73}\) These studies report the viscous filament break-up time to be \( t_b = k^{\lambda \eta C_a} \), with \( 6 < k < 33 \). This time scale results from the competition between surface tension trying to shrink the liquid filament and the viscous stresses in the filament opposing it.

Incorporating the expressions for \( t_{sq} \) and \( t_b \) into Eq. (16) we obtain

\[
T_{sq} = \frac{d_C}{U_C} [1 + k \lambda Ca]. \tag{18}
\]

Given the volumetric contribution to the drop size during the squeezing period is \( V_{sq} = T_{sq}Q_d \), we have

\[
\frac{V_{sq}}{hw_d^2} = \beta Q + \xi Ca D, \tag{19}
\]

where \( \xi = k^{\lambda \eta C_a} \).

Solving Eqs. (14), (15), and (18) together gives the final prediction for the generated droplet volume in T-junction devices taking into account the viscosity ratio of the fluids.

In Figure 10, we compare the predictions of our improved model with the simulation data across a wide range of viscosity ratios and for two geometries with different aspect ratios \( H = 0.33 \) and 1. For the first geometry, \( H = 0.33 \), the viscosity ratio varies between 0.01 and 15 and for the second geometry, \( H = 1 \), we have \( \lambda = 0.1 \) and 10. Two sets of fit parameters \( \xi \) and \( \beta \) were used for these two geometries. We find excellent agreement between our model predictions and the simulation data capturing different viscosity ratios. For each set of data with the same \( Ca \), we found the best-fit values of \( \xi \). For the first geometry, \( H = 0.33 \), this value varies between 7 and 36 for the entire set of data with an average of \( 21.96 \pm 8.6 \). For the second geometry, \( H = 1 \), the best-fit value of \( \xi \) varies between 8 and 19 with an average of \( 14.8 \pm 3.69 \). These finding are consistent with the range of \( k \) values reported in the literature (5 < \( k \) < 33) since we have \( w_{hil} \sim O(1) \). According to McKinley and Tripathi, the \( k \)-value depends on the longitudinal stress in the viscous thread, which is a function of the axial curvature of the filament.\(^{69}\) In our study, the rate at which the dispersed phase is injected into the droplet changes when the capillary number is varied. This can cause the axial curvature of the dispersed phase finger to vary with the capillary number and therefore it can result in the variability of \( k \)-values, as observed when we generated the parity plot (Fig. 10). We also found the best fit parameter \( \beta = 1.18 \pm 0.4 \) for the first geometry, and \( \beta = 1.4 \pm 0.77 \) for the second geometry. These values are reasonable estimates since we get \( \beta = 1.8 \) and 1.15 for the first and second geometry, respectively, when evaluated with the theoretical model proposed by van Steijn et al. that includes the effect of T-junction geometry.\(^{31}\)

FIG. 10. Parity plot showing the prediction of the improved model incorporating viscosity ratio, versus the simulation data for (a) \( W = 1, H = 0.33, \) and 0.01 \( \leq \lambda \leq 15 \) and (b) \( W = 1, H = 1, \) and \( \lambda = 0.1 \) and 10. In this plot, \( V \) is the volume of the droplet normalized by \( hw_d^2 \). Model predictions reported by Garstecki et al. and van Steijn et al. are shown by horizontal solid and dotted lines, respectively.\(^{30,31}\) The grey lines indicate 10% deviation from the expected true value. The insets show the model predictions (dotted lines) as a function of \( Ca \) for different \( Ca \) values: \( \square—0.003, \bullet—0.005, \circ—0.008, \square—0.01, \) and \( \Box—0.015 \).
An important outcome of our analysis is that it explains why at a given (low) capillary number, the generated drop volumes are the same for \( \lambda < 1 \) but varies for \( \lambda > 1 \) (see Fig. 8). As shown in Fig. 11, when we plot the contributions of \( t_{sq} \) and \( t_b \) to the overall squeezing time as a function of the viscosity ratio, we find that for \( \lambda < 1 \), the contribution of viscous thread breakup time is negligible compared to the squeezing time. This implies that the produced droplets will have the same size for \( \lambda < 1 \). Alternatively, we observe that for \( \lambda > 1 \), the breakup time contributes more making the drop volumes dependent on the viscosity ratio.

IV. CONCLUSIONS

We implemented the volume-of-fluid (VOF) method for investigating droplet generation in microfluidic T-junction devices. We examined the different regimes of the dispersed phase behavior and found the VOF simulations to be in good agreement with experimental maps of the regimes. We also performed simulations at low and high capillary numbers, as well as in different T-junction geometries. The predicted droplet size was in good agreement with experimental reports in the literature. Taken together, these results validate our VOF methodology and choice of simulation parameters.

Our simulation results show that for \( \lambda > 1 \), the droplet size depends on the viscosity ratio. Previous theoretical models have not considered this dependence of the viscosity ratio on the generated droplet size. We developed an improved model and incorporated this dependence by including the break-up time needed to fragment the high-viscosity dispersed phase. This improved model predicts successfully the influence of drop volume on the capillary number as well as the viscosity ratio.

More broadly, we find that VOF is a useful tool for droplet-based microfluidics and has the potential to reveal new insights into the fluid physics of drop formation and behavior, which can have important technological applications.

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