A Novel Fluid-solid Coupling Framework Integrating FLIP and Shape Matching Methods

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ABSTRACT

Physically-based fluid animation and solid deformation driven by numerical simulation have manifested their significance for many graphics applications during the past two decades. For example, the fluid implicit particle (FLIP) method and shape matching technique based on position based dynamics (PBD) have demonstrated their unique graphics strength in fluid and solid animation, respectively. We propose a novel integrated approach supporting the seamless unification of FLIP and shape matching. We devise new algorithms to tackle existing difficulties when handling new phenomena such as high-fidelity fluid-solid interaction and solid melting. The key innovation of this paper is a unified Lagrangian framework that seamlessly blends FLIP and PBD based shape matching constraint towards the natural yet strong coupling between fluid and deformable solid. Within our integrated framework, it enables many complicated fluid-solid phenomena with ease. We conduct various kinds of experiments. All the results demonstrate the advantages of our unified hybrid approach towards visual fidelity, efficiency, stability, and versatility.

CCS CONCEPTS

•Computing methodologies →Physical simulation;

KEYWORDS

FLIP, Shape Matching, Position Based Dynamics, Solid Deformation

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1 INTRODUCTION

Physically-based simulations of fluids and deformable solids have been widely studied in graphics. In particular, some applications involve many interesting phenomena related to fluid-solid coupling and interactions, such as solid motion and deformation, fluid coupling with rigid/soft bodies, etc. However, certain difficulties still prevail and call for novel algorithms and techniques. In this paper, we mainly focus on the complicated dynamic interactions between fluids and solids, offering a novel method for the complicated fluid-solid phenomena that can not be realized by naive combination of certain methods. We examine the fluid implicit particle (FLIP) method, which is widely used to simulate high-quality fluid effects because of its less numerical dissipation and better numerical stability [20]. Our improved solution is the unification of FLIP and position-based dynamics (PBD) to streamline fluid-solid coupling in a single framework.

The state-of-the-art methods for visually-appealing fluid animation can be roughly categorized into grid-based and particle-based approaches. Grid-based approaches usually partition the simulation domain into grid cells and calculate the physical properties at each grid cell, such as the Lattice Boltzmann method. In contrast, for the capture of sub-grid details, particle-based approaches are more flexible than grid-based ones. For example, Both PBD [10] and smoothed particle hydrodynamics (SPH) methods [14] do great works. In recent years, FLIP becomes a very popular particle-grid coupling method, which is good at handling incompressible fluid with complex boundaries [5–7]. Although many great ideas have been proposed for fluid simulation based on FLIP or FLIP-coupled methods, FLIP based fluid-solid interactions and two-phase fluids animations have not been studied as widely as SPH and PBD methods.

In contrast, PBD methods are commonly used for solid simulation with high-level stability. In this paper we will extend and unify the incompressible FLIP method and PBD method [4, 13] to uniformly accommodate multiple phases with ease, including deformation bodies, fluid-solid coupling and solid melting. Of which, the distributions of all phases (fluids and solids) are uniformly represented by FLIP particles. The dynamics of the multi-phase system are governed by the shape matching constraints, which serve as constraint conditions of PBD method. Our salient contributions can be summarized as follows:

- We propose a coupled FLIP-shape matching framework, which enables to simultaneously simulate a much wider range of fluid-solid phenomena.
- We propose to uniformly model the behaviors of all the involved materials based on the same set of variables in FLIP Navier-Stokes equations, which greatly reduce the computation complexity.
- We propose new boundary handling method for solid and penetration prevention measurement, which ensure the stable and robust simulation.

2 RELATED WORK

Since this paper mainly focuses on fluid-solid interactions, to keep the review most relevant to our work, we briefly summarize previous works as follows.

FLIP-based fluid simulation. FLIP method is introduced to computer graphics by Zhu et al. [20], and then it is extended to simulate splashing water [7, 8], preserve fluid sheet [2], conduct fluid-solid coupling [15], combine with particle methods [6], model multi-scale droplet/spray [19], etc. For example, Ando and Selino et al. [2, 6] respectively proposed methods to improve the particle distribution of FLIP method. Boyd and Bridson [5] extended FLIP method to model two-phase flows, named as MultiFLIP, which separates velocity fields with a combined divergence formulation to enforce overall incompressibility.

PBD based simulation. PBD is used to handle position-level constraints based on iterative Gauss-Seidel solver [4, 12]. Many works employ PBD for the simulation of deformable objects. For example, Müller et al. [13] introduced a geometric constraint to PBD for deformable object simulation, which serves as the basic framework of our solid simulation. Bender et al. [3] proposed a continuum-based train-energy formulation to solve the constraint function. Tournier et al. [17] formulated a compliant constraint to avoid instabilities due to linearization, which enables the unification of elasticity and constraints. Meanwhile, PBD is also extended for fluid simulation. Macklin et al. [10] proposed a set of positional constraints to enforce constant density, and then they handled the contact and collision of fluid-solid particles in a unified manner [11], which is flexible enough to model various materials.

In summary, compared to the aforementioned works, we blend FLIP and PBD based shape matching constraint towards the strong coupling between fluid and deformable solid, using FLIP particles for all materials instead of PBD particles. In contrast to PBD particles, FLIP particles have communication among themselves, and the particles solely communicate via the underlying grids [7]. That is the very reason we wish to introduce FLIP into our unified framework.

3 INTEGRATED FRAMEWORK

Both FLIP and PBD methods can satisfactorily simulate a variety of scenes, however, for some complicated phenomena such as fluid and deformable solid interaction, they fail to provide a pleasurable and convincing result without further improvement. The key of our approach is a unified Lagrangian framework that blends FLIP and PBD based shape matching constraint via natural and strong coupling. At the numerical simulation level, we take advantages

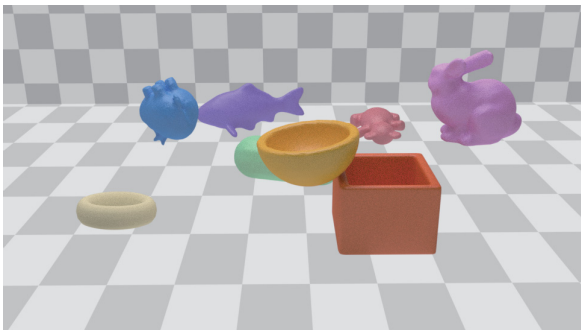

Figure 1: Elastic bodies' dynamics simulation based on our method. All the particles are computed by FLIP algorithm and the shapes are maintained by shape matching of PBD.

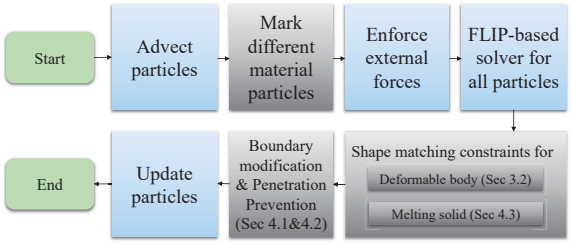

Figure 2: The pipeline of our integrated framework. The blue squares represent the normal FLIP algorithms, and the gray ones represent our improvements being coupled in FLIP framework.

of MultiFLIP and shape matching. The dynamics of particles are solved by MultiFLIP solver, and the deformation of solid is handled by shape matching constraint. Our hybrid framework consists of three main components: (1) the uniform solution of all particles in a FLIP framework; (2) the coupling of FLIP simulation and shape matching constraint; (3) the correction of particles to ensure the accuracy and stability of FLIP solver. And the pipeline of our integrated framework is shown in Fig. 2, which illustrates it can support seamless unification of FLIP and shape matching.

3.1 The Unified Algorithm

Take fluid-solid interaction for example, Algorithm 1 shows the main process within a time interval Δt. The texts marked in blue highlight our method's contributions, which improve the standard MultiFLIP simulation for hybrid fluid-solid simulation. During initialization, we mark different materials with different flags (e.g., F_{solid},F_{fluid}). FLIP solver provides two positions ($\mathbf{x}_\mathbf{p}^0$ and $\mathbf{x}_\mathbf{p}$) for each solid particle, which can respectively be used as the original and predicted position for shape matching constraint. So we can apply the shape matching constraint to solid particles directly to simulate solid bodies movements (Fig 1).

Algorithm 1 Implementation of our integrated framework for fluid-solid simulation.

1: Advect velocities of particles
2: Enforce external forces (gravity)
3: Verify fluid and solid particle flags by F_{solid},F_{fluid}
4: Map all particles to grid $\mathbf{u}_g^0 \leftarrow \mathbf{u}_p^0$, $\mathbf{x}_g^0 \leftarrow \mathbf{x}_p^0$
5: Compute level set Φ and velocity on grid $\mathbf{u_g}$
6: Project $\mathbf{u_p} \leftarrow \mathbf{u_g}$, $\mathbf{x_p} \leftarrow \mathbf{x}_g^0 + \mathbf{u_g}\Delta t$
7: **if** particle $\in F_{solid}$ **then**
8: Project shape matching constraint
9: Compute target position $\mathbf{g}$
10: Update $\mathbf{x}_p^*$+=$\alpha\ (\mathbf{g} - \mathbf{x_p})$
11: Update $\mathbf{u}_p^* \leftarrow (\mathbf{u_p}, (\mathbf{x}_p^* - \mathbf{x}_p^0)/\Delta t)$
12: **else** Continue
13: Correct boundary conditions
14: Perform mechanism of penetration prevention
15: Update the velocities and positions of all particles
16: Update particles' flags

3.2 Integrated Formulations

When getting the particle velocity of solid, we compute a predicted position for this particle via

$$\mathbf{x}_p = \mathbf{x}_p^0 + \mathbf{u}_p \Delta t, \tag{1}$$

and $\mathbf{x}_p^0$ is the initial position, and $\mathbf{u}_g$ is the projected velocity from FLIP solver.

The target position is computed via

$$\mathbf{g} = C_{sm}(\mathbf{x}_p, \mathbf{x}_p^0) \times (\mathbf{x}_p^0 - \mathbf{x}_{cm}^0) + \mathbf{x}_{cm}, \tag{2}$$

with predicted mass center $\mathbf{x}_{cm}^0$, and $C_{sm}(\mathbf{x}_p, \mathbf{x}_p^0)$ is the shape matching constraint related to the initial and predicted positions. Thus, new particle position is computed as:

$$\mathbf{x}_p^* = \mathbf{x}_p^* + \alpha(\mathbf{g} - \mathbf{x}_p). \tag{3}$$

The computations in line 10 and line 11 of Algorithm 1 are the same as the traditional PBD method, wherein $\mathbf{x}_p^*$ will be updated toward the final position of each particle. And we update the new velocity $\mathbf{u}_p^*$ by a combination of FLIP-velocity $\mathbf{u}_p$ and displacement-velocity:

$$\mathbf{u}_\mathbf{p}^* = \frac{\left(\mathbf{u_p} + \left(\mathbf{x}_\mathbf{p}^* - \mathbf{x}_\mathbf{p}^0\right)\right)/\Delta t)}{2}. \tag{4}$$

With the hybrid framework, we can simulate fluid interactions with various materials ranging from stiff solids to viscoplastic bodies. However, since the new values are directly computed based on a PBD constraint, which involves no restrictions related to boundaries and FLIP fluid, the fluid particles may penetrate into solid, or the solid particles may move out of the defined boundary. Thus, boundary conditions need be added to solid and fluid particles to guarantee the stability, and the penetration prevention measurement should be introduced to guarantee physical reality and accuracy.

To track the sharp interface between different materials, we respectively use two sets of independent matching-cube algorithms to capture the fluid and solid (or some other materials') surfaces. And we use the boundary particles of the solid to sample the surface of objects [1], which allows handling different shapes, such as lower-dimensional rigid bodies. The flag of each particle indicates which system it belongs to, ensuring the accurate demarcation of different materials.

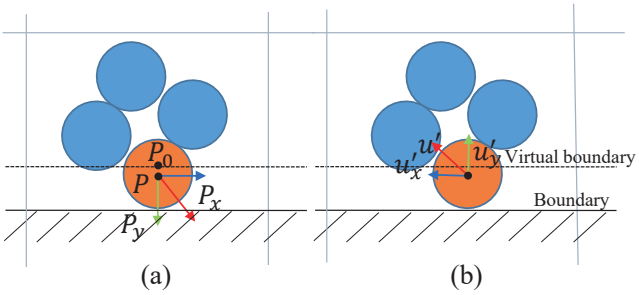

Figure 3: Illustration of solid boundary conditions. When a solid particle penetrates into the virtual boundary (dotted line), we give it an inverse velocity depending on the position P and the position of virtual boundary grid $\mathbf{P}_0$.

4 BOUNDARY HANDLING AND IMPLEMENTATION DETAILS

With our hybrid framework, we can simulate fluid flows and imitate a solid simulator with shape matching constraint. However, since the naive coupling will suffer from serious numerical and stability problems [9], we will introduce our novel measurements into the hybrid framework to realize accurate simulations and rich applications.

4.1 Boundary Handling for Solid

In our hybrid solver, fluid grid interacting with boundary grid will rebound in an inverse direction. But for solid, since a set of particles are clustered together, if we take the same boundary conditions as fluid, local movements of the particles on boundary grids will lead to unrealistic global deformation, and affect the simulation stability. Thus, we define a new boundary condition for solid. For a set of solid particles, we allow transitory penetration into a virtual boundary grid to keep the global shape unchanged. The virtual boundary grid is defined as:

$$\begin{cases} \mathbf{P}_{min} = \mathbf{X}_{min} + \lambda h, \\ \mathbf{P}_{max} = \mathbf{X}_{max} - \lambda h. \end{cases} \tag{5}$$

Here $\mathbf{X}_{max}$ is the maximum position of the boundary grid, and $\mathbf{X}_{min}$ is the minimum position of the boundary grid, $\mathbf{P}_{min}$ and $\mathbf{P}_{max}$ are the virtual boundary locations, and h is the grid size of FLIP. $\lambda \in [0, 1]$ controls the shrinkage degree of virtual boundary. When $\lambda = 0$, the virtual boundary is equal to real boundary. In most of our experiments, we set $\lambda = 0.3$, which can effectively avoid penetrating into the real boundary and can handle the solid interactions well.

As shown in Fig. 3(a), $\mathbf{P}$ is the position of solid particle that penetrates into virtual boundary, and $\mathbf{P_0}$ is the position of the virtual boundary grid. When solid particle moves across the virtual boundary, we compute its inverse velocity (shown in Figure 3(b))

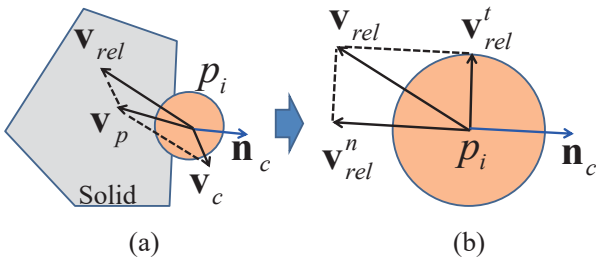

Figure 4: Penetration prevention for particles. $\mathbf{v}^n_{rel}$, $\mathbf{v}^t_{rel}$ are the relative velocities along the normal and tangential directions.

by

$$\mathbf{u}' = \begin{cases} \mathbf{u}^* + \gamma(\mathbf{P}_0 - \mathbf{P}) \times N, & \mathbf{P} < \mathbf{P}_{min} \\ \mathbf{u}^* - \gamma(\mathbf{P} - \mathbf{P}_0) \times N, & \mathbf{P} > \mathbf{P}_{min} \end{cases}. \tag{6}$$

Here γ is a bounce parameter that can be taken as the boundary elasticity, and N is the total number of solid particles. Because solid particles' number relates to the mass of the solid body, the more particles a solid body has the larger inertia it carries.

4.2 Penetration Prevention Measurement

When simulating fluid-solid interactions with particle-based method, an indispensable work is to prevent fluid particle from penetrating into solid particles. Inspired by the position correction idea used for smooth interface [5], we develop an additional algorithm to prevent particle penetration. First, after updating the particles' positions and velocities, we detect collision between solids and fluid particles. For each fluid particle P_i, we search solid particles that collide with P_i, and determine the collision positions. Second, we compute the relative velocity between the fluid and the solid particle by $\mathbf{v}_{rel} = \mathbf{v}_p - \mathbf{v}_c$, as shown in Fig. 4(a). When the relative velocity points into the solid, i.e., $\mathbf{v}_{rel} \cdot \mathbf{n}_c$ is negative, penetration occurs. To prevent it we impose a fluid-solid boundary condition on the velocity of fluid particle $\mathbf{v}_p$ as:

$$\mathbf{v}^{new}_p = \mathbf{v}_p - \mathbf{v}^n_{rel} = \mathbf{v}_p - (\mathbf{v}_{rel} \cdot \mathbf{n}_c)\mathbf{n}_c, \tag{7}$$

where the relative velocity along the normal direction $\mathbf{v}^n_{rel}$ is subtracted to make the particle's velocity equals to the solid's velocity in the normal direction.

In fact, the above velocity correction on fluid particle is equivalent to enforcing an impulse on the particle. To conserve momentum during the collision handling, we compute the collision force and impose this force on the solid via

$$\mathbf{f} = \frac{m_p(\mathbf{v}_{rel} \cdot \mathbf{n}_c)\mathbf{n}_c}{\Delta t}. \tag{8}$$

4.3 Melting Simulation

Each particle is attached by a material flag in order to identify which material it belongs to, and and the phase change process will be in charge of the flag updating, which also serves as the criterion that guides us to design the dynamic algorithm accordingly. So for each particle, we update its temperature in each time step, and then determine whether its phase needs to be changed or not.

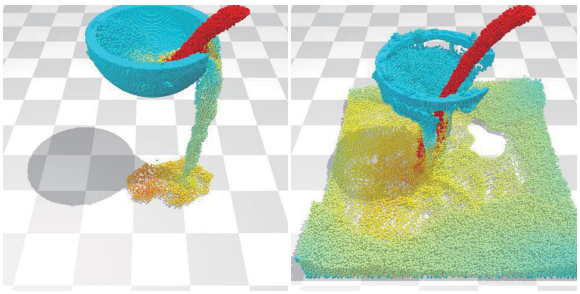

Figure 5: Illustration of heat transfer among solids and liquids. The particles are colored according to temperatures (blue means low and red means high).

Table 1: Performance of experiments.

Scenes	Total particles	Grid size	Avg. time /Timestep(ms)	Rendering time /Frame(s)
Fig. 1	$390k$	64^3	79.33	12.95
Fig. 6	$80k$(fluid)	64^3	82.06	9.75
Fig. 7	$680k$	$128^2 \times 64$	138.68	17.67
Fig. 8	$60k$(fluid)	64^3	71.32	7.37

To simulate melting, we initiate all the particles with a temperature attribute, the temperature update only depends upon the heat transfer. In each time step, temperature is mapped from particles to each grid cell with a weighting function:

$$\frac{T^{n+1} - T^n}{\Delta t} = b\left(\frac{\partial^2 T^{n+1}}{\partial x^2} + \frac{\partial^2 T^{n+1}}{\partial y^2} + \frac{\partial^2 T^{n+1}}{\partial z^2}\right), \tag{9}$$

where T^n is the given temperature field obtained in the last time step, T^{n+1} is the current one we need to update, and b is the corresponding thermal diffusivity parameter. After updating the temperature on grid T^{n+1}_g, we map the temperature changes of the grid to particles, and then update the particle temperature by $T^{n+1}_p = T^n_p + \Delta T_p$. When the temperature of a solid particle reaches to melting point, convert it to a fluid particle and alter it with the fluid's attributes, then release it from being confined as solid. Thus, this particle will become a free fluid particle, while the other solid particles still hold the integrity of solid constraints. Fig. 5 shows the heat transfer process.

5 EXPERIMENTS AND EVALUATIONS

We implement our method on a PC with an NVIDIA Geforce GTX 1080 GPU, Intel Core I7 CPU using c++ and CUDA. And we demonstrate the capabilities of our hybrid framework via several simulation scenarios. Table 1 documents the performance of our experiments, indicating the high efficiency of our CUDA-based implementation.

5.1 CUDA-based Numerical Computation

We implement the entire modeling framework based on CUDA for efficiency. For each particle, we invoke a CUDA thread to calculate which grid cell it belongs to, and then use a CUDA thread for each grid cell to interpolate its velocity from particles. Table 2

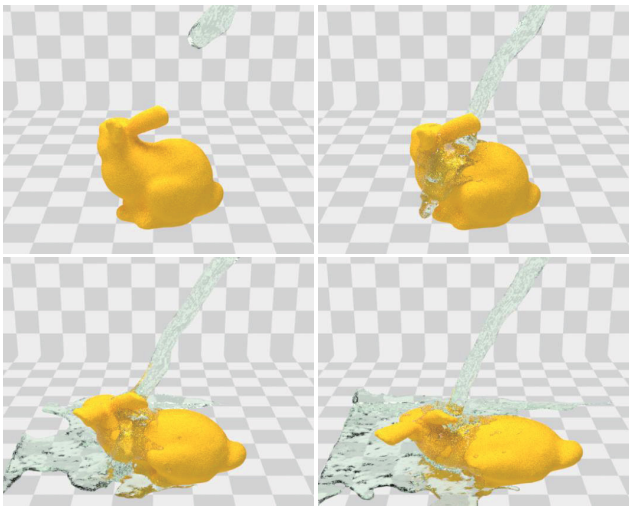

Figure 6: Deformation. From top to bottom: Fluid pours on a deformable bunny.

Table 2: Time performance comparison (in milliseconds (ms)).

Method	Particles		Avg. Time/Timestep (ms)
	Fluid	Solid	
[18]	5000	-	21.32
[16]	5000	23095	62.47
Ours	5000	8199	25.56

documents our method's comparison with [16, 18]. As shown in Table 2, the comparison is based on a scenario that fluid particles pour into an elastic cone. To couple fluid and soft body while avoiding penetration, the method [18] costs less time because of its simple solid structure, but its visualization and applications are far less effective than ours. And when compared to [16], our method uses fewer particles while still achieving better performance. Please refer to our supplementary video for more vivid animations.

5.2 Graphics Results and Discussion

Rigid and elastic bodies. Fig. 1 shows the simulations of elastic bodies' dynamics. Each body has a shape matching constraint, and all of its attributes are solved by MultiFLIP solver. This scenario illustrates that, with our hybrid framework, we can simulate most types of solid dynamics that traditional PBD could accommodate.

Deformation. Fig. 6 demonstrates the realistic rendering results of plastic body deformation. As the water flows over the surface of the plastic bunny, solid particles are enforced to change the global topology, which is handled by the shape matching constraint.

Fluid-solid interaction. Fig. 7 shows a scenario of fluid-solid interaction. A bunny toy interacts with dam-breaking water, which deforms and twists under the force of wave. To imitate the buoyancy force of solid, we add an external force (with inverse gravity direction) to each solid particle. The buoyancy needs to be smaller than gravity, so that the solid particle will float on the water when interacting with fluid particles.

Melting. Fig. 8 demonstrates a fluid-solid coupling and melting scenario (Fig. 5 provides the particles' display), wherein the ice melts due to the heat absorption from the hot water. When a solid particle's temperature rises to melting point, it is removed from shape matching constraint, and becomes a fluid particle. This scenario exhibits the capability to simulate local melting phenomenon, which cannot be achieved by traditional PBD methods.

Limitation. Even though the coupling of FLIP and shape matching models, together with custom-designed algorithms, enables more flexibility to empower simulation results, one limitation of our method is the non-conservation problem pertinent to our improved shape matching constraint. Although the numerical dissipation can be ignored in the vast majority of cases, it expects to deteriorate as the complexity of details increases. For some complicated phenomena such as splash and turbulence, we will have to consider tradeoff between preserving the details and ensuring the stability of the algorithm.

6 CONCLUSION AND FUTURE WORK

We have detailed a novel physically-based framework for fluid-relevant phenomena simulation by integrating FLIP and shape matching constraint. The proposed new algorithms can well combat the existing difficulties in accommodating new phenomena, such as high-fidelity fluid-solid interaction, solid melting, and two-phase immiscible fluid animation. The novel technical elements include boundary handling algorithm, penetration prevention measure, melting model, and improved constraints for two-phase fluid interaction. We have illustrated various types of experiments and demonstrated the advantages of our unified framework.

At present, our integrated framework has already successfully simulated numerous fascinating scenes, there are still many interesting constraints that could be chosen from, some more complex constraint extensions such as cluster-based deformation, lattice shape matching may support more applications. We shall continue to expand our constraints to enable more complex visual-fidelity applications in the future.

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Figure 7: Fluid-solid interaction. From top to bottom: Fluid interacts with a deformable bunny.

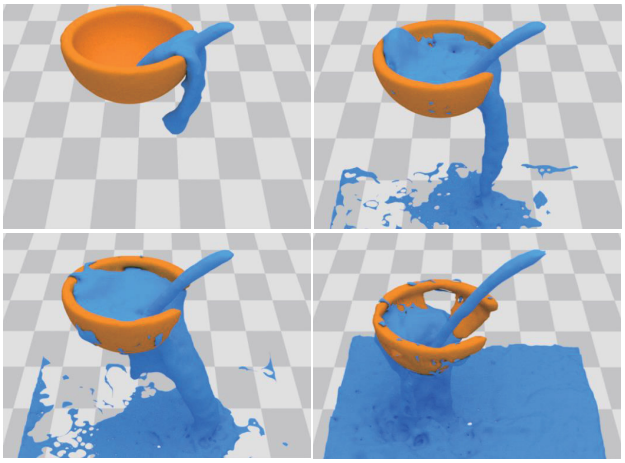

Figure 8: Melting simulation. Fluid pours on a meltable bowl, and the bowl continuously melts.

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