Robust and Efficient SPH Simulation for High-speed Fluids with the Dynamic Particle Partitioning Method

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Abstract

In this paper, our research efforts are devoted to the efficiency issue of the SPH simulation when the ratio of velocities among fluid particles is large. Specifically, we introduce a k-means clustering method into the SPH framework to dynamically partition fluid particles into two disjoint groups based on their velocities, we then use a two-scale time step scheme for these two types of particles. The smaller time steps are for particles with higher speed in order to preserve temporal details and guarantee the numerical stability. In contrast, the larger time steps are used for particles with smaller speeds to reduce the computational expense, and both types of particles are tightly coupled in the simulation. We conduct various experiments which have manifested the advantages of our methods over the conventional SPH technique and its new variants in terms of efficiency and stability.

CCS Concepts

\textsuperscript{•}Computing methodologies \rightarrow Animation; Physical simulation;

1. Introduction and Motivation

Most of the current SPH models are either using a unified constant time step, or adopting an adaptive one which is directly derived from the CFL condition by using the maximum velocity of all particles. To address the efficiency issues in turbulent fluid simulation, we propose a dynamic particle partitioning SPH method (PPSPH) that improves the simulation’s efficiency through changing the time step size for different fluid particles. We divide each simulation time step into certain numbers of small sub-steps, and in each sub-step we simplify the computational procedure when computing the particles with low velocities, and then promise a correct final result at the end of each major step. Specifically, the salient contributions of this paper include:

\begin{itemize}
  \item We introduce a k-means clustering scheme into SPH to partition particles based on their velocities.
  \item We propose a dynamic two-scale time step model for particles with large difference in velocities.
  \item Our method is easy to be performed on most of the existing SPH frameworks and can be implemented on GPU for high-efficiency applications.
\end{itemize}

Compare to the tradition SPH framework, our simulation step consists of three main components: (1) Partitioning particles into two disjoint groups by k-means clustering; (2) Uncoupling high-speed particles from fluids and simulating faster particles with refined time-step size; and (3) Integrating all fluid particles including faster ones and slower ones.

2. Related Works

In recent years, a lot of visual-effect-dominant fluid simulation methods have been introduced for visually-appealing animation. Fluid implicit particle (FLIP) method [ZB05] is a popular particle-grid coupling method to simulate multiphase fluid [BB12, GLY\textsuperscript{\ast}17], splashing water [GB13, YLHQ15], fluid-solid coupling [GLQH17], etc. Also, data-driven methods [RWTT14, LJS\textsuperscript{\ast}15] and position based fluid [MM13, MMCK14] all achieve good achievements. However, among the fluid simulation methods, SPH may be the most widely used and most studied method.

The SPH method is commonly used in computer graphics for fluid simulation [Mon92], and is extended for free surface flow [Mon94]. Müller et al. [MCG03] derived inter-particle forces from Navier-Stokes equations for interactive water simulation, but it was hard to model strong incompressible scenes. The fluid simulation method of Müller et al. [MCG03] was based on equation of state to weakly enforce incompressibility. Later, WCSPH [BT07] was proposed to enforce a weakly compressible form of SPH for fluid flow based on the Tait equation. Implicit pressure solvers

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Figure 1: A brief illustration of simulation steps. There are mainly three parts of a major step, the first one includes a pre-process that partitions particles into two disjoint groups, the last one integrates all particles with a refined time step for both types of particles, and the rest mini steps only simulate high-speed particles.

was introduced by Solenthaler et al. in PCISPH [SP09]. Ihmsen et al. proposed IISPH [ICS’14]. And Bender et al. [BK17] employed two solvers to enforce a constant density condition and a divergence-free condition. And there have been some asynchronous methods [HF17] [TPS08] [GB14] proposed to boost the efficiency of SPH simulations. Ihmsen et al. [IOS’14] summarized the state-of-the-art research within the graphics community.

3. Particle Classification

The function of k-means clustering is partitioning n points into k clusters. We use predicted particle velocities as sample observations to partition particles into two different clusters. The predicted velocity is computed as

\[
v_i^* = v_i + \Delta t \frac{F_{\text{ext}}}{m_i}.
\]

The k-means clustering usually converges to a local optimal value. In order to have a more clear dividing line for the particles, our method usually sets the maximum predicted velocity to the initial mean value of one cluster, and sets the threshold value to another cluster. The threshold value is a user-defined value used to determine whether the k-means clustering procedure will be proceeded. In the end of the clustering, particles will be partitioned into two clusters, one with slower particles, another with faster ones. And please note that, the clustering and the time-step updating procedure are only performed in the first mini step of every major time step.

4. Simulation Steps

4.1. Major Step

A major step consists of \( N_{\text{mini}} \) mini steps, but the major time-step size is not a simple multiple of mini time-step size in most cases during the simulation. Particles with high velocity will be processed in each mini time step, and we should guarantee that each particle has a normal motion so we implement the mini and the major time step framework in the simulation. According to the CFL condition

\[
\Delta t \leq 0.4 \frac{d}{\|v_{\text{max}}\|_2},
\]

apply the maximum velocity of all particles \( v_{\text{max,all}} \) and the maximum velocity of particles in low-speed group \( v_{\text{max,low}} \) to Eq. (2) we get the major and the mini time step size

\[
\Delta_{\text{mini}} \leq 0.4 \frac{d}{\|v_{\text{max,all}}\|_2} \quad \Delta_{\text{major}} \leq 0.4 \frac{d}{\|v_{\text{max,low}}\|_2}.
\]

Also, we set a user-defined threshold (\( N_{\text{tres}} \)), when the ratio is smaller than the threshold, the major step will only contains one standard SPH time step. To prevent the ratio from being too large which may cause an unreal simulation result, we set a maximum mini time step amount \( N_{\text{max}} \). The amount of mini step in a major step is computed as

\[
N_{\text{mini}} = \begin{cases} \min(N_{\text{max}}, \frac{\Delta_{\text{major}}}{\Delta_{\text{mini}}}) & \text{if } \frac{\Delta_{\text{major}}}{\Delta_{\text{mini}}} \geq N_{\text{tres}}, \\ 1 & \text{else} \end{cases}
\]

where we set the threshold \( N_{\text{tres}} \) in Eq. (4) to a number less than 1.5. After determining the amount of mini steps \( N_{\text{mini}} \), different strategies will apply on the following mini steps.

4.2. Mini Step

In our method, during the first \( N_{\text{mini}} - 1 \) mini steps, only particles partitioned into the high-speed group will process pressure solving and update their velocities and positions, because we have noticed that the pressure solving procedure is the most time-consuming procedure in the whole simulation step. And in the last mini step, all particles in the scene will process pressure solving and update their positions.

In the last mini step, we process the complete SPH time step procedure for all particles in the scene, but with some modifications. Since during the first \( N_{\text{mini}} - 1 \) steps, low-speed particles are waiting for the high-speed particles, to make up the difference in these two-scale time steps, we propose a solution of having an adjustment on the time step for low-speed particles:

\[
\Delta t_i = \begin{cases} \varepsilon \Delta_{\text{mini}} & \text{if particle } i \in \text{low-speed cluster} \\ \Delta_{\text{mini}} & \text{else} \end{cases}
\]
Table 1: Performance of k-means clustering.

<table>
<thead>
<tr>
<th>Number of Particles</th>
<th>350k</th>
<th>550k</th>
<th>1.2m</th>
<th>3.0m</th>
</tr>
</thead>
<tbody>
<tr>
<td>k-means Clustering</td>
<td>29.1ms</td>
<td>40.04ms</td>
<td>64.4ms</td>
<td>138.6ms</td>
</tr>
<tr>
<td>Time Step Update (per mini step)</td>
<td>8.48ms</td>
<td>10.5ms</td>
<td>29.1ms</td>
<td>79.0ms</td>
</tr>
</tbody>
</table>

Table 2: Performance comparison for fluids flushing scene.

<table>
<thead>
<tr>
<th>Method</th>
<th>Pressure Solving</th>
<th>Overall Simulation</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>avg</td>
<td>sum</td>
</tr>
<tr>
<td>PCISPH</td>
<td>223.5ms</td>
<td>14454s</td>
</tr>
<tr>
<td>PP-PCISPH</td>
<td>114.2ms</td>
<td>918s</td>
</tr>
<tr>
<td>DFSPH</td>
<td>161.8ms</td>
<td>1036s</td>
</tr>
<tr>
<td>PP-DFSPH</td>
<td>71.5ms</td>
<td>55s</td>
</tr>
</tbody>
</table>

This scene contains 359,000 fluid particles in the fluid block with 100,000 high-speed particles to emit later and 448,000 static boundary particles.

where \(\varepsilon\) is a user-defined parameter of time compensation of low-speed particles in the last mini step. The parameter \(\varepsilon\) is usually set to a number larger than 1. The higher value \(\varepsilon\) is, the more temporal details in the simulation will be produced and more time will be spent. And \(\varepsilon\) should not set to a number larger than a value the CFL condition indicates. In our implementation, we usually set the number to 1.5 to ensure the equilibrium of detail description and the efficiency when \(N_{\text{max}} \leq 3\).

5. Experimental Results and Discussion

In this paper, we integrate our method with both PCISPH and DFSPH framework, and employ the cubic spline kernel in simulation and use XSPH variant [SB12] to compute viscosity terms for low viscosity fluids, and employ the method proposed in [AIA∗12] to handle the boundary. We fully implement the simulation procedure of our methods on CUDA for efficiency.

Table 1 compares the efficiency of k-means clustering procedure in our method (PP-SPH) with the time step updating procedure of standard SPH methods. Note that, in this measurement, k-means clustering also includes the time step computing procedure. Although clustering cost some time, we process the clustering procedure only once per \(N_{\text{mini}}\) steps. It has obvious advantage when simulating high-speed fluid, since the greater the speed difference is, the larger \(N_{\text{mini}}\) will be, and the better performance we could have than the standard ones.

5.1. Comparisons with Different Parameter Settings

We have two more parameters in our method than the standard SPH methods, the maximum mini steps in a major step \(N_{\text{max}}\) and the time step adjustment parameter for low-speed particles \(\varepsilon\).

Table 3 demonstrates the performance with different parameter settings in double dam break scenes. In our method, choosing a larger value for \(N_{\text{max}}\) will improve the efficiency of the simulation. In our simulation, we usually choose a number ranging from 2 to 4 for \(N_{\text{max}}\). The \(\varepsilon\) is served as a supplement to the time step for low-speed particles, a larger \(\varepsilon\) will decrease the efficiency of the simulation as computational cost will increase with the time-step size.
5.2. Large-scale Scenarios

Fig. 3 illustrates a fluid block falling and flushing to the street, hitting the vehicles and street lights. Fluids produce big waves when hitting the buildings and flowing through the roads between buildings. Both scenes have proved that our method is capable of simulating large-scale scenarios with stability and efficiency. We believe that, our method should be able to simulate some larger scenarios, yet due to the limitation on the graphics memory at our lab, we are not able to conduct such simulations currently.

6. Conclusion

In this paper we have detailed a novel SPH framework to boost the efficiency and robustness of fluid simulation. In contrast to the previous SPH methods using a static or adaptive yet unified time step for all fluid particles in the simulation, we introduced a two-scale time-step scheme for SPH, which partitions particles into two groups, and enforced different schedules and different time-step sizes for particles of different groups. Simulating particles with different strategies in fluid simulations has resulted in an improvement of computational efficiency, and can guarantee the numerical stability while still producing realistic details. Our method has apparent advantage over existing SPH methods when simulating large-scale phenomena with high-speed particles in the scenes.

Limitation. We noticed that some particles with intermediate speeds will be rapidly shifting between high-speed and low-speed clusters. Although it won’t affect the stability of our simulation, it might slightly reduce the efficiency of our simulation because more particles are partitioned into the high-speed group, thus more computations need to be handled.

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