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Ray-Traced Collision Detection: Interpenetration Control and Multi-GPU Performance

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Figure 1: 512 concave mesh objects fall on a planar ground, up to 7300 pairs of objects are tested in the narrow phase.

Abstract
We proposed in [LGA13] an iterative ray-traced collision detection algorithm (IRTCD) that exploits spatial and temporal coherency and proved to be computationally efficient but at the price of some geometrical approximations that allow more interpenetration than needed. In this paper, we present two methods to efficiently control and reduce the interpenetration without noticeable computation overhead. The first method predicts the next potentially colliding vertices. These predictions are used to make our IRTCD algorithm more robust to the above-mentioned approximations, therefore reducing the errors up to 91%. We also present a ray re-projection algorithm that improves the physical response of ray-traced collision detection algorithm. This algorithm also reduces, up to 52%, the interpenetration between objects in a virtual environment. Our last contribution shows that our algorithm, when implemented on multi-GPUs architectures, is far faster.

Categories and Subject Descriptors (according to ACM CCS): I.3.1 [Computer Graphics]: Hardware Architecture—Parallel processing I.3.5 [Computer Graphics]: Computational Geometry and Object Modeling—Physically based modeling

1. Introduction
In virtual reality (VR) and others applications of 3D environments, collision detection (CD) is an essential task. Given a set of objects, those which might collide have to be known. Although this formulation is simple, collision detection is currently one of the main bottlenecks of VR application because of the real-time constraint imposed by the direct interaction of the user and the natural complexity (\(O(n^2)\)) of the naive algorithms. Recently, new approaches using GPGPU (General-Purpose computing on Graphic Processing Unit) have emerged while taking advantage of the GPU computational performances [AGA11].

In [LGA13] we presented an iterative ray-tracing algorithm for collision detection that uses spatial and temporal coherency to improve performance. The main idea is to update when possible the result of the previous time step rather than computing new collisions from scratch. When
compared to non-iterative algorithms, the speedup obtained with this algorithm is up to 33 times. Nevertheless, the main drawbacks of this method are: (1) it makes approximations when updating the collision and (2) the overall technique makes approximations on the direction of the physical reaction. These approximations may let the objects enter in further collision, increasing the total amount of interpenetration in the virtual scenes.

In order to make this algorithm usable and efficient, we propose to control the interpenetration without loss of performances. This is performed using two methods that correct the approximations. The first one is a novel collision prediction method that predicts future collisions and re-project the errors when updating the previous result. The second one is a ray re-projection algorithm aiming at correcting the direction of the physical reaction.

We also propose to use multi-GPU architectures to exploit further the parallelization potential of the iterative and the predictive algorithm.

The paper is organized as follows: Section 2 presents the related work. After a brief introduction of our previous algorithm [LGA13], Section 3 explains how we can reduce the amount of interpenetration in the physics simulation with our predictive collision detection and re-projection algorithms. Section 4 shows how we can use several GPUs to improve the performances. Section 5 describes performance improvements in our test scenes. Section 6 concludes this paper and gives future work.

2. Related Work

In this section we focus on the literature of collision detection that is the most relevant to our work. For a deeper introduction we suggest the reader to refer to surveys on the topic [TKH’05, KHI’07]. Collision detection has to deal with a large variety of objects and properties of these objects. All these parameters can change the method used, and thus have created a lot of different solutions. Objects can be concave or convex, with or without holes. Objects can be solid, deformable, liquid and topology may change (they can split or merge). Objects can be represented with polyhedrons (generally with triangular faces), CSG (Constructive Solid Geometry) or parametric functions.

Hubbard [Hub93] proposed to decompose collision detection in two phases, the broad-phase and the narrow-phase. The broad-phase plays the role of a filter by performing a cheap non-collision test to quickly remove non-colliding objects. The narrow-phase works with the pairs of objects obtained from the broad-phase. These pairs are in potential collision and the goal of the narrow-phase is to perform an exact collision test to give contact or interpenetration information so the physical response can be computed.

The broad-phase takes a set of objects as inputs and outputs a set of potentially colliding pairs. Nowadays this phase is less critical than the narrow-phase for performance. Aside brute force, broad-phase algorithms can be classified in three categories: spatial partitioning, kinematic and topology.

The narrow-phase takes the pairs produced by the broad-phase and performs an accurate collision test. It outputs a set of colliding objects and information about the contacts between objects for the physical response. Narrow-phase algorithms can be classified in four categories: feature-based, simplex-based, bounding volume hierarchy and image-based.

Nowadays GPUs (Graphics Processing Units) are broadly used to improve the performance of collision detection. [LHLK10] perform a massively parallel version of sweep-and-prune algorithm on GPU for the broad-phase. [PKS10] execute both broad and narrow phases on GPU and multi-GPU architectures with spatial subdivision.

2.1. Ray-Traced Collision Detection

Ray-Tracing Collision Detection Algorithms belong to the image-based category of narrow-phase algorithms. Several algorithms have been proposed in the literature. Wang and al. [WFP12] launch rays from an irregular grid to compute a layered depth image, the density of rays is higher around small objects to avoid missing them. Cinder algorithm [Kno03] casts rays from the edges of an object and counts how many object faces the ray passes through. If the result is odd then there is collision otherwise not. These rays are cast toward a regular grid.

Hermann and al. [HFR’08] proposed to detect collision by casting rays from the vertices of the objects. The algorithm is placed on the narrow-phase of the collision detection pipeline and thus works on pair of potentially colliding objects. Rays are cast from each vertex of each object in the opposite direction of their normal. If a ray hits inside the other object before leaving the source object then a collision is detected. Figure 2 shows an example. The plain arrows are three rays that spot collision, the dotted arrow is a rejected ray because it hits the other object from outside.

![Figure 2: Hermann and al. algorithm in 2D.](image)

The main advantage of this algorithm is the information given for physical response. For other narrow-phase algorithms, when a collision is spotted additional computation is needed to compute the physical response. Hermann and al. algorithm gives for each penetrating vertex a direction and a
distance to separate the two objects (as can be seen in Figure 2). The problem is, using the rays as is for the collision response in the physical simulation gives incorrect behavior because the rays are not always correctly oriented to give a realistic physical response.

When using repulsion forces, Hermann and al. proposed an algorithm to weight the forces in order to reduce the tangential forces, but they are not suppressed. When using constraint-based response, weighting cannot be used as constraint cannot be weighted. Furthermore to reduce the cost of constraint solver, contact-points reduction is applied on each colliding pair. After the reduction, the selected subset may contain only incorrectly oriented constraint. This will cause an incorrect physical response that can lead to accept further interpenetration by letting the objects sliding into each other.

### 3. Minimum Penetration Control

Based on Hermann et al. algorithm, we introduced in [LGA13] IR TCD: a new Iterative Ray-Traced Collision Detection algorithm that is used to update the collision in a pair of objects when their relative displacement is inferior to a threshold $\text{displacementThreshold}$. At each time step if the displacement exceeds the threshold we compute the rays with a standard algorithm (and reset the displacement to 0) otherwise we update the ray of the previous step with the iterative algorithm. The iterative algorithm relaunches the ray on the previously colliding triangles and iterates over the neighboring triangles until the new colliding ones are found.

In this section we present how we reduce the total amount of interpenetration by making the detection and the physical response better. Section 3.1 explains how we strengthen the iterative algorithm with predictions. Section 3.2 shows how we improve the physical response.

#### 3.1. Predictive Collision Detection

The main drawback of the iterative algorithm is it only works with previously detected rays. When new vertices enter in collision they will be taken into account only in the next standard step. This may postpone the collision detection of a pair of objects for several time steps, this will lead to accept further interpenetration. The first row of Figure 3 shows an example of this situation. At $t = 0$ a standard algorithm is executed and no collision is detected. At $t = 1$ the previous rays are updated, as the two objects were not colliding at $t = 0$ there are no rays to update. In this case the iterative algorithm fails to detect the collision. At $t = 2$ a standard algorithm is executed and the collision is detected. In this example the detection is postponed for only one step, but in practical cases it can be more.

We propose to solve this problem by performing a predictive ray/triangle intersection. When a ray does not detect any collision (and thus the corresponding vertex is not in collision), we cast a predictive ray from the same vertex but in the opposite direction. If this predictive ray hits a triangle and if the distance is short, the corresponding vertex may hit that triangle (or the neighboring ones) in a near future. Equation 1 gives the distance from which we can cull the predictive rays because the distance is too big. $\text{cullingDist}$ depends on $\text{displacementThreshold}$ and a confidence level of at least 1.

\[
\text{cullingDist} = \text{confidence} \times \text{displacementThreshold} \quad (1)
\]

The predictive rays are injected in the iterative algorithm as candidates for the next steps. Algorithm 1 gives the ray/triangle-mesh intersection algorithm that can handle prediction. The modification from the original algorithm [LGA13] is the addition of line 6 and 7. If the ray does not hit the current triangle, before looking at the neighboring triangle, we check if the current triangle is behind the ray. In such case the corresponding vertex is not in collision and the current triangle is kept as a candidate for the next step.

The second row of Figure 3 shows how these predictive rays prevent delayed collision detection. At $t = 0$ a standard algorithm is executed and no collision is detected, predictive rays are cast outside the objects and three predictive rays kept. At $t = 1$ the predictive rays are updated and collision is successfully detected, at $t = 2$ a physical response can be applied to prevent further interpenetration.

We cast rays from the vertices that are inside the intersection of the bounding volumes, this is an optimization proposed by [HFR’08] to reduce the number of rays. In the context of predictive rays this optimization may discard predictive rays and delay the collision detection. To avoid this problem we extend the intersection of the bounding volumes by a distance of $\text{distanceExtension}$ in all the directions.

<table>
<thead>
<tr>
<th>$t=0$</th>
<th>$t=1$</th>
<th>$t=2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>standard</td>
<td>iterative</td>
<td>standard</td>
</tr>
</tbody>
</table>

**Figure 3: Comparison of predictive and non-predictive collision detection, with predictive rays the collision is detected earlier resulting in less interpenetration.**

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Algorithm 1 Iterative ray/triangle-mesh intersection with prediction handling

1: function itRAYTRIINTERSEC(Ray ray, Triangle tri)
2:     for i = 1 → maxIt do
3:         intersection ← cast ray on tri
4:         if ray hits tri then
5:             return intersection
6:         else if opposite(ray) hits tri then
7:             return keepAsPrediction(tri)
8:         else
9:             edge ← closestEdge(tri, ray)
10:            tri ← adjacentTriangle(tri, edge)
11:        end if
12:     end for
13: return intersectionNotFound
14: end function

which corresponds to a confidence zone. distanceExtension must be greater than displacementThreshold, this ensures that we do not discard predictions that are essential for the iterative algorithm. The value of distanceExtension must be minimized for better performances because higher values increase the number of ray cast thus increasing computation.

In term of performance, casting a second ray in the opposite direction theoretically doubles the cost of the standard ray-tracing algorithm. In practical case we use two properties to reduce the cost of the predictive rays. First the rays share several parameters in common, they have the same starting point and follow the same line (in the opposing direction). These shared parameters allow to factorize a portion of the two rays depending on the ray-tracing algorithm used. The second property is that we do not need to follow the predictive ray beyond the distance distanceExtension as it is exiting the confidence zone. This allows to shorten the ray traversal thus making it less expensive.

The evaluation of this predictive algorithm is given in Sections 5.2 and 5.3.

3.2. Ray Re-projection

With Hermann et al. algorithm, the orientation of the rays does not give the optimal direction for the physical response. Figure 4.a shows an example, the rays correctly detect the collision but their directions do not correspond to the needs of the physical response (normal and tangential information) shown in Figure 4.b.

When executing the ray-tracing from the vertices, we record for each ray the normal of the impacted surface. Then we propose to post-process the rays by re-projecting them on the normal of the surface they collide. This re-projection minimizes the length of the rays. In physical simulation it corresponds to the Minimum Translational Distance (MTD) [CC86], which is considered as a good heuristic for a physical collision response. An example of a ray re-projection is shown in Figure 4.b.

Projecting a ray on a vector can be done quickly with Equation 2, where \( \hat{r} \) is the ray (the vector from the ray origin to the impacted point) and \( \hat{n} \) is the unitary normal of the impacted surface.

\[
proj(\hat{r}, \hat{n}) = \hat{n} \times (\hat{r} \cdot \hat{n})
\] (2)

Collision response uses contact points to represent collision. A contact point is generally represented by one collision point \( P \) with its penetration depth \( pd \) and its separating direction \( d \) (an example can be found in Figure 5). With this representation we can avoid to compute explicitly the re-projection of the ray with \( P = \) colliding vertex, \( pd = \hat{r} \cdot \hat{n} \) and \( d = \hat{n} \).

This method makes no assumption on the physical response used and thus can be used with any physical response algorithms.

Figure 4: Collision between the box and the ground. In (a) the rays do not give the optimal separating direction. In (b) the rays have been re-projected on the normal of the surface they hit, these rays give the optimal collision response.

Figure 5: A contact point is composed of a point \( P \), a separating direction \( d \) and a penetration distance \( pd \).

4. Adaptation for Multi-GPU Architectures

In [LGA13] we proposed to execute the iterative and standard ray tracing algorithms on GPU to improve performance. GPU works with highly parallelized execution, we have to send to the GPU the work as an array of threads of several indices. We proposed to work with two indices.
The first index iterates through the pairs of objects from the broad-phase and each pair appears two times in opposite order. The second index iterates through the vertices of the first object of the pair, in each thread we cast the ray from the current vertex of the first object of the pair on the second object. This division on the work allows to parallelize the ray-tracing by executing one ray cast per thread. Figure 6 shows an example of such distribution.

We propose to exploit several GPUs to increase performance. To divide the work between the GPUs we break the first iteration into several blocks and execute these blocks on different GPUs. Figure 7 gives an example with three GPUs. The iterative algorithm uses at each step the result of the previous step, this previous result is held in the memory of the GPU that proceed it. Memory transfers between two GPUs are expensive. To avoid transferring the previous result to another GPU we add a constraint: when a pair of objects is proceeded on one GPU, it must be processed on the same GPU until the pair is removed from the broad-phase.

The experimental scenes were developed using Bullet Physics (http://bulletphysics.org/). The broad-phase and the physical response are executed on the CPU. Our narrow-phase algorithm was developed with OpenCL (http://www.khronos.org/opencl/) and can be executed on one or several GPUs. This setup generates memory transfers between the CPU and the GPU as our algorithm is located in the middle of the physical simulation pipeline. In a real situation, the whole physical simulation pipeline would be implemented on GPU thus removing the memory transfers.
The iterative ray-tracing algorithm needs a standard ray-tracing algorithm as a starting point, we have implemented two algorithms:

**Basic traversal:** This algorithm does not use any acceleration structure, for each ray we iterate through each triangle. This method has a high complexity but is simple.

**Stackless BVH traversal:** This algorithm uses a bounding volume hierarchy (BVH) as an accelerative structure [WBS07] with a stackless traversal for GPUs [PGSS07]. This algorithm is more computationally efficient but has increased memory usage due to auxiliary data structures.

### 5.2. Predictive Detection Performances

To evaluate the importance of predictive rays, we run our two sample scenes with and without the predictive rays. Throughout the simulation we compare the results of the iterative and iterative + predictive algorithms with a standard ray-tracing algorithm used as a reference. Figures 9 and 10 give the percentage of incorrect rays (i.e. when the result between the standard and our algorithm is different). Results show a reduction up to 91% of the percentage of incorrect ray with the collision prediction algorithm.

![Figure 9: Percentage of erroneous ray in the first scene.](image)

To evaluate the impact of the predictive rays on the physics simulation we measure the cumulative length of all rays with a standard algorithm along the simulation, this value measures the amount of interpenetration in the whole scene. In the first scene the cumulative length of rays decreases by 45% when using predictive rays. In the second scene we get a 56% decrease. This mean that the predictive rays effectively reduce the total amount of interpenetration.

### 5.3. Predictive Time Performances

To measure the impact on the computation time of predictive rays, we measure the time taken by the ray-tracing with and without the predictive ray with both ray-tracing algorithms.

![Figure 10: Percentage of erroneous ray in the second scene.](image)

Figures 11 and 12 show the time spent on ray-tracing along the simulation for each scene on a Nvidia GTX 660.

When the iterative algorithm is used (e.g., part C of Figure 11), the usage of predictive rays has a small impact on the computation time. This is because the casting of the predictive ray in the iterative algorithm can be factorized with the casting of the standard ray, resulting in a low overhead.

When there is less spatial and temporal coherency and the standard ray-tracing algorithm is intensely used (e.g., B of Figure 11), the impact on the computation time of the predictive rays depends on the ray-tracing algorithm used.

With the basic traversal we can completely factorize the casting of the predictive with the standard ray to avoid traversing the object twice. This makes the cost of the predictive ray negligible. In parallel the usage of predictive rays reduces the amount of interpenetration in the scene which reduces the number of collision and consequently reduces the number of rays to compute. Finally the cost of casting the predictive rays is lower the gain we have by decreasing the number of rays to compute. This result can be seen in part B of Figure 11 where we have an average 1.27 speedup.

With the stackless BVH traversal we cannot factorize the traversing of the predictive ray with the standard ray. But we can limit the traversal of the BVH by limiting the ray length to a maximum of $\text{cullingDist}$ as explained in Section 3.1 to avoid doubling the cost of the standard algorithm. In our experimental scenes, the worst speedup we have by adding predictive ray to the stackless is an average of 0.78 in part B of Figure 11.

These results show that the computational cost of using the re-projection algorithm is low, in the worst case we have a 0.78 speedup and in the best case we have a 1.27 speedup.

### 5.4. Ray Re-projection Performances

As a test we can consider the case of an object in collision with the ground, in such case the reaction should be orthog-
Figure 11: Time spent executing ray-tracing with and without predictive rays in the first scene.

Figure 12: Time spent executing ray-tracing with and without predictive rays in the second scene.

Figure 13: To test the average deviation of the reaction of an object and the ground, we perform the collision detection with several interpenetration distance and orientations.

Table 1: Average deviation of penalty forces in percent for three meshes against the ground.

<table>
<thead>
<tr>
<th>Technique</th>
<th>sphere</th>
<th>bunny</th>
<th>pig</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of vertices</td>
<td>70</td>
<td>453</td>
<td>1085</td>
</tr>
<tr>
<td>Standard</td>
<td>3.97</td>
<td>22.0</td>
<td>27.0</td>
</tr>
<tr>
<td>Weighting [HFR’08]</td>
<td>2.66</td>
<td>16.5</td>
<td>20.6</td>
</tr>
<tr>
<td>Re-projection</td>
<td>0.00</td>
<td>0.00</td>
<td>0.00</td>
</tr>
</tbody>
</table>

The deviation of the forces with the sphere is low. This can be explained by the regularity of the shape. Due to the symmetry, the forces balance with each other. With irregular objects forces do not balance as well and deviation is not negligible. The weighting technique proposed by Hermann et al. reduces the tangential forces, but it only reduces the deviation by about 25% for irregular shapes. Our re-projection technique guarantees a 0% deviation as the forces are reoriented toward the optimal direction.

We have tested our re-projection algorithm with our two experimental scenes, the collision response is managed with a sequential impulse solvers with contact-points reduction. We perform the simulation with and without ray re-projection and measure the total length of the rays along the simulation as a global interpenetration measurement. Table 2 gives the cumulative length of all rays for both scenes with and without ray re-projection and the decrease percentage. Results shows that our solution effectively reduces the total interpenetration, up to 59%.

<table>
<thead>
<tr>
<th></th>
<th>Standard</th>
<th>Re-projection</th>
<th>Decrease</th>
</tr>
</thead>
<tbody>
<tr>
<td>Scene 1</td>
<td>16 115</td>
<td>10 793</td>
<td>59%</td>
</tr>
<tr>
<td>Scene 2</td>
<td>66 094</td>
<td>38 398</td>
<td>42%</td>
</tr>
</tbody>
</table>

Table 2: Comparison of the cumulative length of all rays along the simulation with and without ray re-projection.

5.5. Multi-GPU Performances

We run our experimental scene with the iterative ray-tracing algorithms on one Nvidia GTX 580 and on multi-GPU setup of two Nvidia GTX 580. Figures 14 and 15 show the time spent on ray-tracing on the GPU at each simulation step for each scene. With two GPUs, we get an average speedup of 1.77 and 1.66 for respectively the first and second scene.

6. Conclusion and Future Work

In the context of iterative ray-tracing for collision detection, our collision prediction algorithm avoids to miss new collision when the iterative ray-tracing algorithm is used. Results show a reduction of the total interpenetration in our test scenes up to 56% with our collision prediction algorithm. We also presented a ray-reprojection algorithm to improve the
collision response. Total interpenetration in our test scenes has been reduced up to 52%. We have presented an adaptation for multi-GPU architectures that gives a speedup up to 1.77 with two GPUs.

In our methods we have three constants along the simulation: \textit{maxIt} and \textit{displacementThreshold} that comes from the iterative ray-tracing algorithm, and \textit{distanceExtension} that come from the collision prediction algorithm. Their values depend on the object sizes and velocities in the scene. In future work it would be interesting to study more deeply the impact of the choice of these values on the simulation and to try to use dynamics values instead of constants in order to optimize these values along the simulation.

We also want to study the multi-GPU scalability of our algorithm when using many GPUs to be able to simulate more complex scenes.

It also would be interesting to extend our algorithm to deformable bodies. With rigid bodies we can use accelerative ray-tracing structure with a high construction cost. In the case of deformable bodies the ray-tracing accelerative structure needs to be updated at each time step making complex ray-tracing techniques more expensive. In addition self collision has to be detected. We believe that our iterative, predictive and re-projection algorithms are suitable for deformable bodies and should improve the performance of ray-tracing collision detection algorithms for deformable bodies.

References


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