



This presentation is a follow-up of our HPG paper that we presented just a couple of days ago.

The paper is called "Concurrent Binary Trees for Large-Scale Game Components"...



…and most of the content of the paper is shown here, including…



4 pages of implementation details shown in red!

The reason I'm showing these is that it turns out that since we wrote the paper 6 months ago, Anis re-worked the implementation to make it even faster.



So what we describe in the paper is now more or less deprecated already!



And so in this presentation Anis will share all the details of his new implementation.

Before he does so, I will quickly recap what the paper is about to bring everyone up to speed.



The paper introduces a new algorithm to deal with large-scale game components.



A large-scale game component is typically what makes the virtual world of your game look "big".



Taking a few video games as example, that would be terrains, …



…, oceans, …



… or entire planets.

In each of these screenshots, pretty much everything except the characters in the foreground is rendered using a dedicated system.

This dedicated system is what we refer to a "large-scale game component".



Most of the time, such components occupy a lot of pixels so it's important to have a set of efficient algorithms to render them as fast as possible.

The goal of our paper is to contribute to this set of efficient algorithms, …



…which can typically be classified into two categories.



First there's the data-generation category, which addresses how to generate textures, sprites, instances, etc.



Second the triangulation / rendering category, which focuses on how to render this data.



Our paper contributes only to the latter (so we won't be discussing data-generation here) with a triangulation method capable of handling very large environments that can be explored at any different scales.



And actually, our contribution is an improvement over something called "Concurrent Binary Trees" (CBTs), which we presented in the same course 3 years ago.

I like to refer to our improvement as "CBT version 2", or simply CBT-V2. I will explain what CBTs are in just a minute.



Before I do just that, here is a look at what CBT-V2 can render.

Let's have a look at a video that captures the result of our method, which runs at 250+FPS on a PS5 level hardware at full HD resolution.



To provide a better sense of scale of what our CBTV2 produces, we are going to have a look at how dense the triangulation of this shot is.



Here is an alternative view of the same shot and we are going to zoom into it until we reach the resolution of the mesh



















In terms of numbers, it turns out the full resolution mesh of our Earth model, i.e., without any form of level-of-detail as we do here, would require exabytes of data.

An exabyte is a million terabytes, so it would not even fit in a large SSD.



This prevents the use of LOD systems like Nanite, which requires the full resolution mesh as input.



Note that in UE5.3, you could still handle very large meshes using their very own large-scale game component, which consists in coupling two representations: one for close scale, and the other for far away scales.

Unfortunately, hybrid representation are hard to use especially with free-flight cameras because it becomes really tricky to set the location of the transition between both representations.

A nice advantage of our method is that it relies on a single representation so you don't need to worry about these issues.



Right, now I am going to quickly explain what a concurrent binary tree (CBT) is and how our method works with them, then Anis will dive into the details.



A concurrent binary tree of CBT is a full binary tree (so each node has exactly two children except for the leaves) with two main parts.



The first part is a bitfield located at the bottom of the tree.

So the leaf nodes only store binary values.



The second part is a sum-reduction tree of this bitfield.

So all remaining nodes store the number of green bits, i.e, bits set to one, in its corresponding subtree.



This means that the root node gives the total number of green bits.

In addition, the sum-reduction tree makes it possible to iterate over the green bits, even if the green bits aren't located sequentially in the bitfield.


These properties makes it possible to use a CBT as memory pool manager that tracks allocated and available memory.

A memory pool is simply an array of whatever data you want to store and we set its capacity to that of the bitfield.

We then track allocated entries using green bits and available memory with red ones.

This way the root of the CBT gives us how much memory is available / allocated.



We can then implement a simple allocation and de-allocation operator as follows.



For allocation we set a bit to one and update the sum-reduction tree.



For de-allocation we set a bit to zero and again update the sum-reduction tree.



Now here is what we do for our triangulations.



Our method takes a halfedge mesh as input…



… and produces a triangle for each halfedge (feel free to go back and forth between this slide and the previous one).



We then compress the 3 vertices of each triangle into a single integer value that we call a heapID and store it in a dedicated entry of the memory pool.

In this example we have 12 triangles so we require 12 slots in the memory pool.



As an example, the triangle 23 is stored in slot 7.

In addition to the heapID, we also store neighborhood information with pointers.



Continuing with triangle 23, the neighbors are triangle 24, which is located at slot 8…



… triangle 27, which is located at slot 11 …



… and triangle 17, which is located at slot 1.



And naturally we do this for each triangle.



The reason we store this information is because we implement a bisection scheme that can split triangles into two new ones.

Naively bisecting a triangle would produce a T-junction, which would result in cracks in the final surface.



But thanks to the neighborhood information, we can propagate bisections across multiple triangles to guarantee crack-free surfaces.



Last thing for me: the memory pool we use for our demo is 128k wide, which requires 7 MB of memory in total.

That concludes my overview of the paper.

The key takeaway here is that CBTs provide a way to allocate, release, and iterate over all the elements of a memory pool.

And as Anis will show, all this can be done efficiently on the GPU.

Your turn Anis!



We're going to dig into some of the implementation details that allowed us to reach reasonable performance numbers with this method.

That said, we'll not go into code details due to the limited time we have.

The full source code of the demo is released and is available for you to explore and play with it!



First, We'll precise the implementation of the CBT.

We'd like to store the CBT in the group shared memory to make the access more efficient when reading and writing.

As an example, we'll use the 128k elements CBT in this presentation, but as we mentioned before we can go higher (or lower) depending on the needs of the application

The CBT needs to be readable and writable from worker threads. We could use uint32\_t to benefit from the atomic intrinsics, but there is an issue with that:



The number of nodes of the CBT is twice the size of the bitfield, which would, if naively stored would be around one 1MB



However, the group shared memory storage is limited to 32KB on dx12 so we need a better representation!



Each level of the tree is defined by 4 things:

- The number of nodes within the level
- The range of values each node can represent
- The minimal size of each node to represent that range
- The final size used to represent each node

So what we'll do is decompose the tree into multiple parts with different constraints



The first part in red that is an atomic-friendly subtree

During modification, all threads can write safely to any node of the tree.

Each node is represented by uint32\_t to benefit from the atomic intrinsics. This is stored in the group shared memory;



Iin green, we have several subtrees, each subtree will only be modified by one thread at a time.

The size of each node is rounded to the closest power of two to represent the data underneath. This is also stored in the group shared memory;

## **Parallel Update Routine (CBT Implementation 128K)**



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Then there are what we call virtual levels. These are not represented explicitly,

## **Parallel Update Routine (CBT Implementation 128K)**





but can be deduced from the last level



Which is the rawbitfield. This bitfield it is represented using uint64\_t to benefit from some intrinsic functions and is stored in a structured buffer outside of the shared memory.



And this the memory footprint of a 128k bit CBT. In the implementation, you'll find the layout for the other sizes of the CBT



Now let's look in practice what that maps to:



We have two structured buffers:

- The first one stores uint32\_t and contains the red and green parts of the tree
- The second stores uint64\_t and contains the bitfield (in blue)



In addition to that we need an allocation buffer which is a uint32\_t buffer, we'll see how it is used in a second



Using the memory footprint table on the right, we can deduce the size of the tree



And that defines the group shared memory space our CBT occupies. In this case the CBT is about 3 KB, which we can consider reasonable.



The way the memory manager is used every frame is the following.

- First we're gonna book the worst case memory we need
- Then we're gonna allocate a bunch of bits, that will depend on the subdivision case we're processing,
- We'll finally return to the memory manager memory space what we didn't consume



The booking and cancelling is done using these functions



And this function that allows us to allocate the next available memory slot



Then once all the allocations have been done, we're gonna decode the location of the bits that we've allocated


To do so, we first have to load the CBT into shared memory for our searches, we do it using this function, this will copy per workgroup the tree buffer into the group shader memory



Then we actually operate the tree descents taking advantage of the very specific memory layout of the CBT which helps us to accelerate significantly the operation.



Then we'll set atomically some bit to 1 (for the allocations that we've done) and 0 for the bit's we've deallocated.

It is done using this function that will operate directly on the bitfield buffer.



And finally we'll operate a sum reduction on the tree to be able to return to a valid state. We can then take advantage of the memory layout we defined to get it performant, but i'll cover it a tad later.



And we'll repeat this operation every frame and that defines how the CBT works as a parallel memory manager



On the other side, we need to specify what is the memory pool like and how it's used



As mentioned before, we need two buffers to store the HeapID and Neighbors



Then we also need to be able to store the camera relative position. Obviously, there are the three positions of the vertices of the triangles



But we actually need to store a 4th position which allows to rebuild the parent triangle and is required for split/merge decisions



In addition to that, we need some temporary data, that we'll call those update data



At each update loop, a bisector can be in one of 5 states, unchanged, single split, double splits (two cases) or triple split.



We encode that split state into three bits and store it in a unit32\_t to use the atomic intrinsics



Depending on the split, we'll need up to 3 news slots to store the location of produced bisectors.

In this implementation, we always reuse the current bisector slot for performance and compactness reasons



The way the indices are assigned has to be consistent



In that way, for any split combination at an interface between two bisectors



We're able to predict the neighbor data using simply the update and neighbor data of the current and neighbor bisectors.

This is important and it is the cornerstone of the sync free parallel implementation



In addition to that we need an array to store additional flags and neighbors propagation data



And with that, we've defined the CBT implementation and the Memory pool layout.

Now let's look at the update routine itself, each block on the right maps to one or multiple compute shader



So for a given triangulation, there is a corresponding memory pool



Most of the passes will be an indirect dispatch based on the a given set of bisectors.



The set of bisectors will change based on the pass



and the subset will be specified at each pass of the pipeline



Alright, the first step of the pipeline is "Reset"



This pass is a single thread dispatch that:

- prepares the allocations for the frame
- resets the bisector queues for indirect dispatches



For the second pass, "Classify", each thread will run for one active bisector



The idea is to classify each bisector in one of 3 states, Keep, Merge or Split. Depending on what we're trying to achieve the criteria for these operations can change, but for the demo here are the ones we used



First, we do a NdotV test and flag it for merge if it fails



We then intersect one or multiple frustums and flag it for merge if it does not intersect any of them



Then we'll project the 4 vertices and evaluate the projected area of the triangle and the parent triangle



We compare those area to a triangle size that is defined by the application and flag the bisector for merge, split or keep depending on the result



Then we move on to the "Split" pass, this will only run on the bisectors that have been tagged for as requiring a split.

In the example that would be only the bisector with the heapid 46.



The first thing we need to do is figure out what is the worst case allocation size we need to guarantee the propagation of this split.

If we cannot guarantee the required memory space, the split cannot go through, that would break the LEB scheme



The naive solution that is 2 \* depth - 1 ends up causing under tessellation issues when moving at a high speed and this is not viable



There is a better way of doing it that we'll not detail in the presentation but you can find the details in the source code we share

Once we have that estimation, we'll try to book that amount of memory from the CBT. Again we fail to guarantee that amount, this split won't go through as we cannot preserve the LEB scheme



If we manage to book the worst case, we de-recursify the propagation algorithm into a while loop that will split each triangle, allocate one or multiple slots, propagate the subdivision to the twins all the bits we need until we're done.



So in this case we do the first allocation, but we have a T junction so we continue on our path


Which leads us to doing two more allocations and we're good to go.



Finally we're return to the CBT the memory space that we didn't allocate



The next step "Decode", runs on the bisectors that have been flagged for a subdivision, (46 and 27 in this case),

This will associate each allocated slot with a free bit of the CBT



For this pass we need to load the CBT into the shared memory first

The number of bits set to 1 of the subdivision pattern gives us the number of decodes we need to do.

We do those decodes and store them in the update data

Once this is done, we move on to the next step



The following pass, "Bisect", that runs on the same set of bisectors than the decode pass, aka every bisector that will be splitted



Based on the subdivision pattern of the current bisector and and the one of its neighbors, we'll modify the heap ID and adjust neighbor pointers.

This only partially updates the neighbors and we still have problematic interfaces (marked in red in the example) that will be resolved in the following pass



This will produce the new state of the triangulation



And finally we'll update the bits of the CBT based on the number of allocations that were required for this split



The next step, "Propagate Bisect", will process these problematic interfaces by changing the neighbors when required



The patching routine has to take into account if the neighbor was split (and how it was split, once, twice or thrice) and adjust the neighbors accordingly

That covers it for the split part of the algorithm. Now we need to process the merges, It is roughly the same approach as the splits, but simpler as there are no propagation to account for.



The first pass is "Prepare simplify", This dispatch will run on the 4 bisectors that have been flagged for simplification



To avoid synchronization and racing conditions, the bisector with the smallest heapID in the four is in charge of checking and registering for the simplification.

In the example, that would be the bisector with the heapID 42

(Actually in practice, we don't enqueue the bisectors with odd heap IDs into the merge queue as an optimization)



We can easily identify which one it is using the depth of the bisectors and their heapIDs



That bisector is then in charge of making sure they all required a merge operation and if it is the case, it will enqueue for the next step



Which is the "Simplify" step, this will only run for the bisector 42 in this case



We update the bisectors that will remain (the ones with the smallest heapIDs for each pair, 42 and 48 in this case) and free the one other ones (43 and 49).



The same way than for split, we update the heapID and only partially the neighbors that will be updated in the following pass. This generates the new triangulation.



Then we set the bits in the CBT to zero for the unused slots



In this last step of the modification "Propagate simplify", we dispatch on the bisectors that generated an inconsistency in the neighbors



And depending on if the neighbors were merged we or not, the routine needs to take account if the bisector was deleted, etc



And with that we've processed our split and merge requests and have a coherent and LEB compatible triangulation



But we're not done yet, there are couple steps left before we get to the end of our update routine.

We did modify the bit field in the previous steps, but the tree now doesn't match anymore. The next one is operating a sum reduction on the CBT's tree



Let's again take the example of the 128k CBT



First we do a reduction to move from our bitfield to the first explicitly stored level of the tree



Then another pass within the pet-thread atomic part of the tree



Finally one single workgroup will do the full sum reduction on the atomic friendly tree. With that we're able to do the have the full tree updated at a reasonable cost.



To illustrate what i mean by reasonable, here you can see a table that recaps the reduction time for each CBT size that we provide in our demo and for 3 GPUs (Intel Arc 770, AMD 6650 XT and Nvidia 4090)



The next step is processing the modified bisectors to ensure that the positions are up to date



As we mentioned before, the geometry information is encoded using the heapID and if we break it down to a binary representation, it would look something like this



The final position is obtained using a chain of matrix multiplications, each bit of the heapID will correspond to one of two matrices



These are the two matrices we use for the demo



There are two issues here:

- First, even if the output position are stored using simple precision floating point and camera relative, we have to do the multiplications in double space (otherwise we break the floating points due to precisions issues due to to our method)

- Depending on the subdivision level, we can have up to 64 3x3 matrix multiplications to evaluate which is a lot



The first thing we can do is split this multiplication into two 32 3x3 float multiplications with a 3x3 double matmul at the middle which reduces the ALU consumption and better uses the FLOPS of the card, but is still not good enough



If we decompose the heap ID into sequences of bits of equal size



The way we can accelerate this even more is by building what we call a matrix cache, it is a premultiplication of all combinations over a certain number of levels (in practice, we've measure that a depth of 5 is the best speedup/size compromise because we need to load this into shared memory as we'll be accessing it quite a bit).



By doing this, we reduce the number of matrices to up to two sets of 6 3x3 mat muls and one double 3x3 matmul which allows us to get very reasonable execution times as you'll see in a second.

This shader executes only for modified bisectors which is a fraction of the total bisectors


In practice, we would load the matrix cache into shared memory



We would flatten the heapID into 4 positions (3 for the current bisector and 1 for the parent)



And then we would apply a transformation to map this to a spherical planet and convert it to relative world space



Finally to all existing positions, we would apply the relevant deformation, for the earth it would be the result of a bunch of FFTs and the moon we use the displacement maps that the NASA provides on it's website, plus a bunch of noise functions for high frequency details



These produced positions would then be used for the rendering rasterized or ray traced



But also for the following frame to operate the classification



And with that, we are ready to render our mesh to screen



The last thing we would like to cover is the performance numbers of the method, this is a screenshot of a capture of the demo running at the surface of the earth



In our case, we're interested by this section, highlighted in red



This table recaps the timings for each of the passes that we've explained. We did the profiling on three GPUs:

- The Arc 770
- the AMD 6650 XT which is roughly equivalent to a PS5
- A higher end GPU, the Nvidia 4090



The important bit there is that we're able to stay at reasonable cost at all time and less than 0.2ms on a hardware equivalent to a PS5 which is more than reasonable given the achievements of the method



The rest of the pipeline is dependent on the actual deformation that we'd use and the general structure of the rendering pipeline.

For the demo we're using a visibility buffer followed by a material pass.



These are numbers we got for the AMD 6650 XT, but these are really dependent on how you generate your surface data really.



In addition to that i'd like to share some number ray tracing related. The method is compatible with real-time hardware ray tracing and given that the number of primitives remains quite "low", we're able to get something decent even by doing a full rebuild every frame of the tlas and blas



And that's it. The full code of the demo that we've show is open-source and can be found in the github that is linked here.

Also, don't hesitate to check the paper for more detailed explanations about the method, in the meantime, we are available for questions