Surface Area Heuristic

The Surface Area Heuristic (SAH) of a split is the potential cost of traversal and intersection for a given split. The SAH itself provides no way of finding the minimal costs. Luckily the minima can only be found on minimum and maximum bounds of an object within the voxel being split. This is because the change in cost is linear between each edge bound of an object. It is then easy to sweep across all possible minima with only O(n log n) complexity. While there are faster algorithms, the time that it would have taken to implement them would not have been worth it for the minimal speed increases that they would have given. Recently the practice of using a favouring function (λ) to increase the chance of a split where one of the sides has no objects has become quite common showing consisent speed increases.

$$\lambda(P_L, P_R, N_L, N_R) = \begin{cases} 80\% & (N_L \equiv 0 \lor N_R \equiv 0) \land (P_L \neq 1 \land P_R \neq 1) \\ 100\% & \text{otherwise} \end{cases}$$

$$C(P_L, P_R, N_L, N_R) = \lambda(P_L, P_R, N_L, N_R)(K_T + K_I(P_L * N_L + P_R * N_R))$$

Algorithm 1: Surface Area Heuristic.

Algorithm 2: Classify.

```
(T_R, T_L) classify (tree, T, p, N_L, N_R, N_P)
           T_L = null; T_R = null;
           T_L = \text{malloc}(N_L + N_P);
           T_R = \text{malloc}(N_R + N_P);
           i_{TL} = i_{TR} = 0
           for(t < T)
              is_left = is_right = false
              for(j = 0; j < 3; j++)
11
                t_b = t_{j,p_k};
13
                 if(t_b < p_b)
                   is_left = true;
                 if(t_b > p_b)
                    is\_right = true;
19
20
              if(\neg is\_left \land \neg is\_right) // planar
21
22
23
                 if(p_{Side} \equiv RIGHT)
24
                    T_R[i_{TR}++] = t;
25
26
                    T_L[i_{TL}++] = t;
27
28
              if (is_left)
29
                 T_L[i_{TL}++] = t;
30
              if (is_right)
31
                 T_R[i_{TR}++]=t;
32
33
           return (T_R, T_L);
34
35
```

Algorithm 4:

```
Node gen_node(tree, V, T, depth)
          Node n;
          (N_L, N_R, N_P, p, side) = find_plane(tree, V, T);
          n_p = p;
           if (depth \equiv tree_{max\_depth} \lor p_{cost} > K_I |T|)
             n_T = T;
              return n;
13
           Voxel V_L, V_R;
14
           voxel_split(p,V, \&V_L, \&V_R);
15
          (T_R, T_L) = \text{classify}(tree, T, p, N_L, N_R, N_P);
16
17
          n_{left} = ge\_node(tree, V_L, T_L, depth + 1);
          n_{right} = ge\_node(tree, V_R, T_R, depth + 1);
18
          return n;
```

Algorithm 3: Find Split Plane. O(n log n)

```
(N_L, N_R, N_P, p, side) find_plane (tree, V, T)
2
          best\_cost = \infty; best\_p = null; side = null; E = null;
          best_N_L = best_N_R = best_N_P = 0;
          for(k = 0; k < tree ->k; k++)
             E = \text{malloc}(2|T|);
10
             for(t \text{ in } T)
11
                 Voxel B;
12
13
                 voxel_gen_from_tri(\&B, t);
                 voxel_clip(\&B, V);
14
                 if(voxel_is_planar(B, k))
15
16
17
                   E[j++] = \{t, B_{min,k}, k, PLANAR\};
18
                 else
19
20
                   E[j++] = \{t, B_{min,k}, k, START\};
21
22
                   E[j++] = \{t, B_{max,k}, k, END\};
23
24
              sort(E); //explain later
25
             N_L = N_P = 0;
26
27
             N_P = |E|;
              for(i = 0; i < |E|)
28
29
30
                p = E[i];
31
                P_{START} = P_{END} = P_{PLANAR} = 0;
32
                 while (i < |E| \land E[i]_b \equiv p_b \land E_{type} \equiv END)
33
34
                \{P_{END}++; i++;\}
35
36
                 while (i < |E| \land E[i]_b \equiv p_b \land E_{type} \equiv PLANAR)
37
                 \{P_{PLANAR}++; i++; \}
38
39
                 while (i < |E| \land E[i]_b \equiv p_b \land E_{type} \equiv \text{START})
40
                \{P_{START} + +; i + +; \}
41
                N_P = P_{PLANAR}; N_R -= P_{PLANAR}; N_R -= P_{END};
42
43
44
                sah_{-}data = SAH(k, p_b, V, N_L, N_R, N_P);
45
46
                 if (sah\_data_{Cost} < best\_cost)
47
48
                   best\_cost = sah\_data_{Cost};
49
                   best_p = p;
                   best\_side = sah\_data_{Side};
50
51
                   best_N_L = N_L; best_N_R = N_R; best_N_P = N_P;
52
                N_L += P_{PLANAR}; N_L += P_{START}; N_P = 0;
54
55
56
           return (best_N_L, best_N_R, best_N_P, best_p, best_side);
57
```

Persistent Threading

Persistent threading is a relatively new algorithm that takes advantage of a rather old strategy. It takes advantage of the improved performance of Warp synchronous execution (a warp is Nvidia's unit for a group of processors that run using SIMT, also known as a Wavefront on AMD hardware). Persistent threading enforces that only a single task (thread) is assigned to each core of the GPU. This allows for improved SIMT performance between cores of a warp/wavefront as it enforces equal distribution of work tasks (especially when there is a non-trivial workload). It gets around the limited workloads of warp synchronous programming by implementing a global work queue that each thread pulls off of. Another issue that persistent threading alleviates is unbalanced workload distribution that can happen with algorithms that don't fit the SIMD model well (such as tree traversal). The GPU scheduler can sometimes distribute work incorrectly where certain warps will be doing all of the heavy lifting for an operation, but because each core only has one thread, persistent threading can bypass the schedulers as the workload must be evenly distributed. This can greatly improve the performance of these algorithms as it takes advantage of the entire GPU.

Short stack k-d tree traversal.

This is one of the fastest methods of k-d tree traversal. It takes advantage of both the Persistent Threading algorithm and the Short Stack k-d tree traversal algorithm. The short stack k-d tree traversal algorithm is a method for traversal that uses the best parts of k-d restart (with pushdown modification) and traversal algorithms with stacks by using a bounded stack. It will keep descending the tree (and pushing down the root) until the traversal algorithm finds a split where both nodes intersect the ray. The algorithm with then disables push down and pushes the farther node to the stack. It will then continue traversing down the closest node. If no intersection is found along the closest path it will then pop the first split off the stack (the next closest to the origin) and traverse. If the stack becomes full, and then doesn't hit anything while it is being emptied, the algorithm will resort to use k-d restart with the aforementioned pushdown modification where it will then restart right before the first split.

Algorithm 5: Persistent Short Stack k-d Tree Traversal.

```
void update_state(tree_buffer, index, *type, *node, *leaf)
  *type = *(uint8*)(tree_buffer+index); // first value of both node and leaf is the type
  if(*type == LEAF)
    kd_serialized_leaf sleaf = *(kd_leaf*)(tree_buffer + index); //need to use serialized types for this
    leaf->type = *type;
leaf->tri_n = sleaf.tri_n; //number of triangles
    leaf->tri_offset = index+sizeof(kd_serialized_leaf);
  else //NODE
    *node = *(kd_node*)(tree_buffer + index);
(index,t,u,v) traverse (ray_buffer, indices, vertices, tree_buffer)
 blocksize_x = STREAM_PROCESSORS_PER_SIMT_GROUP;
 blocksize_y = SIMT_GROUPS_PER_STREAM_MULTIPROCESSOR;
 x = SM\_ID \% blocksize\_x; //Id within the SIMT GROUP
 y = SM_ID / blocksize_x; //Id of the SIMT GROUP within the Stream Multiprocessor
  //NOTE: shared memory is called local memory in OpenCL
  shared volatile next_ray_array[blocksize_y]; //shared across all processors in the multiprocessor
  shared volatile ray_count_array[blocksize_y];
  //NOTE: In the implementation, the warp_counter is initialised on the cpu and copied to the GPU.
  global volatile warp_counter; // global memory is shared accross the entire device.
  next_ray_array[y] = 0;
  ray_count_array[y] = 0;
  (node_ptr, min, max) stack[STACK_SIZE];
 stack_pointer = 0;
 hit\_info = \begin{bmatrix} 0 \\ 0 \end{bmatrix}
  tringle_index = 0;
 t_{min} = t_{max} = 0;

scene_{min} = 0; scene_{max} = \infty;
  kdtree_node root, node;
  kdtree_leaf leaf;
 current, ype = NODE
  pushdown = false;
 rav_index = 0:
   shared volatile int* local_pool_ray_count = ray_count_array+y; //get this SIMT groups ray count shared volatile int* local_pool_next_ray = next_ray_array+y; //get this SIMT groups next ray
   if(x \equiv 0 \land *local_pool_ray_count \leq 0)
     *local_pool_next_ray = atomic_add(warp_counter, BATCH_SIZE); //retrieve and incriment
      *local_pool_ray_count = BATCH_SIZE;
   ray_index = *local_pool_next_ray + x;
   if(ray\_index \ge |ray\_buffer|)
   if(x \equiv 0)
      *local_pool_next_ray += 32;
      *local_pool_ray_count -= 32;
   r = ray_buffer[ray_index];
   if (!collides_voxel(SCENE_V, r, &scenemin, &scenemax))
     scene_{max} = \infty
   stack_pointer = 0;
   root = *tree_buffer:
   while (t_{max} < scene_{max})
     if (stack_pointer = 0)
       node = root;
       current_type = NODE;
      t_{min} = t_{max};
      pushdown = true;
     else //pop a node off the stack
       stack_pointer --:
       svalue = stack[stack_pointer];
      t_{min} = svalue_{min};
```

```
update_state(tree_buffer, svalue_node_pointer, &current_type, &node, &leaf);
100
                 pushdown = false;
101
               while (current_type # LEAF)
103
104
105
                 left\_is\_close = (r_{orig,k} < node_b \lor (r_{orig,k} \equiv node_b \land r_{dir,k} \le 0));
                 first = left_is_close ? nodeleft : noderight;
110
                 second = left_is_close ? noderight : nodeleft;
111
112
                  if(t_{split} > t_{max} \lor t_{split} \le 0)
113
                    update_state(tree_buffer, first, &current_type, &node, &leaf);
117
                    update_state(tree_buffer, second, &current_type, &node, &leaf);
118
119
120
121
                    stack[stack_pointer++] = {second, tsplit, tmax}; // push
123
                    update_state(tree_buffer, first, &current_type, &node, &leaf);
124
                    pushdown = false;
125
126
127
128
                  if (pushdown)
130
                    root = node:
131
132
133
              for(t = 0; t, |leaf_{num_t riangles}|; t++)
134
135
137
                 offset = *(uint*)(kd_tree+leaf_triangle_start+(t*sizeof(uint)));
138
                 for(j = 0; j < 3; j++) //read triangle indices
139
                   tri[j] = read_texture(vertices, read_texture(indices, offset+j)x)xvz;
140
141
142
143
                 if (collides_triangle(tri. &hit_coords. r))
144
                    if(hit\_coords_t \le 0)
145
                      continue:
                    if (hit_coords; < thit)
147
                      thit = hit_coords;
                      hit_info = hit_coordsuv
                      tri_index = offset;
152
154
            result = \{0\};
              result = \{tri_indx, t_{hit}, hit_info_u, hit_info_v, \}
```