A* GPGPU Implementations for Path-Finding

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Overview

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The basics

GPGPU approaches

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The A* search algorithm

- ► A* is a best-first search algorithm
- ► Chooses minimal node f(n) = g(n) + h(n)
- Needs priority queue to select next node
- Needs look-up data structure for visited nodes

Animation CC BY 3.0 Subh83 on wikimedia.org



Multi-agent A* vs. parallel GA*

Problem: It's hard to parallelize the priority queue!

Multi-agent A* [1]

- Per-agent local priority queue
- Node and edge information can be shared. Open and closed list has to be stored for each agent.
- Local memory is a scarce resource! I've seen 16, 32 and 48 kilobyte.

Parallel GA* [2]

- Multiple independent priority queues
- Expand multiple nodes at the same time
- ► The challenge is handling of duplicate nodes!
- Again, bound by memory

Priority queue: Dispatch memory access

```
typedef struct {
   local uint float *localMem;
   const size t localSize;
   global uint float *globalExt;
           size t size;
} OpenList;
uint float read heap(OpenList *open, size t index) {
   return index < open->localSize ?
       open->localMem[index] :
       open->globalExt[index - open->localSize];
}
void write heap(OpenList *open, size_t index,
                                uint float value) {
   if (index < open->localSize)
       open->localMem[index] = value;
   else
       open->globalExt[index - open->localSize] = value;
}
```

Priority queue as binary heap

```
We need the common...
                               But also...
 push
                                 find
 ► top & pop
                                 update
 void push(OpenList *open, uint value, float cost) {
     _push_impl(open, &open->size, value, cost);
 void update(OpenList *open, size t index, uint value,
                                            float cost) {
     push impl(open, &index, value, cost);
```

Binary heap: Push

```
void push impl(OpenList *open, size t *size,
                uint value, float cost) {
    size t index = (*size)++;
    while (index > 0) {
        size t parent = (index - 1) / 2;
        uint float pValue = read heap(open, parent);
        if (cost < pValue.second) {</pre>
            write heap(open, index, pValue);
            index = parent:
        } else
            break:
    _write_heap(open, index, (uint_float){value, cost});
```

Binary heap: Pop

```
void pop(OpenList *open) {
    uint float value = read heap(open, --(open->size));
    size t index = 0:
    while (index < open->size / 2) {
        size t child = index * 2 + 1;
        uint float cValue = read heap(open, child);
        if (child + 1 < open \rightarrow size) {
            uint float c1Value = read heap(open,
                                              child + 1);
            if (c1Value.second < cValue.second) {</pre>
                ++child:
                cValue = c1Value;
        if (cValue.second < value.second) {</pre>
            write heap(open, index, cValue);
            index = child:
        } else break;
```

Binary heap: Find

This could certainly be improved! Now it's basically a simple breadth-first search.

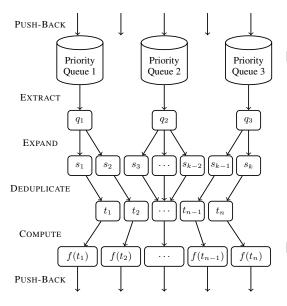
```
uint find(OpenList *open, uint value) {
   for (uint index = 0; index < open->size; ++index) {
      uint_float iValue = _read_heap(open, index);
      if (iValue.first == value)
            return index;
   }
   return open->size;
}
```

Alternative: Except duplicate entries and ignore them later on. Would cost extra memory!

Multi-agent A* kernel

```
push(&open, source, 0.0 f);
while (open.size > 0) {
    const uint current = top(&open); pop(&open);
    /* ... */
    const uint2 range = adjacencyMap[current];
    for (uint edge = range.x; edge != range.y; ++edge) {
        /* ... */
        const uint nblndex = find(&open, nbNode);
        if (nbIndex < open.size &&</pre>
            nbInfo.totalCost <= nbTotalCost) continue;</pre>
        /* ... */
        if (nbIndex < open.size)</pre>
            update(&open, nbIndex, nbNode, nbTotalCost +
                                              nbHeuristic);
        else
            push(&open, nbNode, nbTotalCost + nbHeuristic);
```

Data flow in parallel GA*



Run kernels, in order:

- clearSList
- extractAndExpand
- clearTList
- duplicateDetection
- exclusive_scan
- compactTList
- computeAndPushBack

Image from [2].

Algorithm 1 GA*: Parallel A* search on a GPU

```
1: procedure GA*(s, t, k)
                    \triangleright find the shortest path from s to t with k queues
         Let \{Q_i\}_{i=1}^k be the priority queues of the open list
         Let H be the closed list
        PUSH(O_1, s)
        m \leftarrow \text{nil}
                                         \triangleright m stores the best target state
         while Q is not empty do
             Let S be an empty list
 8:
             for i \leftarrow 1 to k in parallel do
 9:
                 if Q_i is empty then
10.
                      continue
11.
                 end if
12:
                 q_i \leftarrow \text{EXTRACT}(Q_i)
13:
                 if q_i.node = t then
                      if m = \text{nil or } f(q_i) < f(m) then
14.
15.
                          m \leftarrow q_i
16:
                      end if
17:
                      continue
18.
                 end if
19:
                 S \leftarrow S + \text{EXPAND}(q_i)
20:
             end for
21.
             if m \neq \text{nil} and f(m) \leq \min_{q \in Q} f(q) then
22:
                 return the path generated from m
23:
             end if
24:
             T \leftarrow S
25:
             for s' \in S in parallel do
                 if s'.node \in H and H[s'.node].g < s'.g then
26:
27:
                      remove s' from T
28:
                 end if
29.
             end for
30.
             for t' \in T in parallel do
31:
32:
                 Push t' to one of priority queues
                  H[t'.node] \leftarrow t'
33.
34:
             end for
35:
         end while
36: end procedure
```

Run kernels, in order:

- clearSList
- extractAndExpand
- clearTList
- duplicateDetection
- exclusive_scan
- compactTList
- computeAndPushBack

Pseudocode from [2].

Use of synchronization mechanisms: return code

```
kernel void extractAndExpand(/* ... */
                         global uint *returnCode) {
   if (openSize == 0) {
       atomic min(returnCode, 2);
       return; // failure: no path found!
   /* ... */
   if (current == destination) {
       atomic min(returnCode, 0);
       return; // success: path found!
   /* ... */
   atomic min(returnCode, 1); // still running...
```

Use of synchronization mechanisms: inserting elements

```
__kernel void duplicateDetection(/* ... */) {
   /* ... */
    if (nodeInfo.closed == 1 &&
        nodeInfo.totalCost < current.totalCost)
        return; // better candidate already in open list
    const uint hash = current.node % hashTableSize;
    const uint old = atomic xchg(hashTable + hash,
                                 current.node);
    if (old == current.node)
        return; // node has already been added
    global Info *tlist =
        tlistChunks + GID.x * slistChunkSize;
    const uint index = atomic inc(tlistSizes + GID.x);
    tlist[index] = current;
```

Updating information: Transaction approach

```
#pragma OPENCL EXTENSION cl khr int64 base atomics : enable
typedef struct {
    uint closed; uint node;
    float totalCost; uint predecessor;
} Info;
// Update totalCost and pred. as one 64 bit transaction.
ulong *curCostPred = (ulong *) &current.totalCost;
global ulong *infoCostPred =
    ( global ulong *) &info[current.node].totalCost;
ulong oldCostPred = atom xchg(infoCostPred, *curCostPred);
float *oldCost = (float *) &oldCostPred;
// assert: current.totalCost > 0.0f
while (*oldCost != 0.0f && *oldCost < current.totalCost) {</pre>
    // The old entry was better. Swap back!
    *curCostPred = oldCostPred;
    oldCostPred = atom xchg(infoCostPred, *curCostPred);
```

Lessons learned

- OpenCL is hard to make portable and reliable
 - Compiler crashes
 - Wrong work sizes (even through API)
- Performance depends on actual hardware
- CPU A* is hard to beat, random obstacles have a huge impact

Multi-agent A*

- Work items > local memory
- Make use of __private memory?

Parallel GA*

- Needs lots of memory, highly dependent on random obstacles
- Still lots of room for improvement!

Demo configuration on my secondary computer

```
OpenCL device: GeForce GT 520
 ---- CPU reference run...
CPU time for 2500 runs: 0.545225 seconds
 ---- GPU A* run...
GPU time for 2500 runs:
 - Upload time: 0.0184138 seconds
 - Kernel runtime: 1.05191 seconds
 - Download time: 0.00187089 seconds
 ---- CPU reference run...
CPU time for graph (500, 500): 0.655936 seconds
 ---- GPU GA* run...
GPU time for graph (500, 500):
 - Upload time: 0.00613427 seconds
 - Kernel runtimes:
   - CompactTList: 0.0401437 seconds
   - ComputeAndPushBack: 0.129521 seconds
   - DuplicateDetection: 0.102979 seconds
   - ExtractAndExpand: 0.358814 seconds
   - compute::exclusive_scan: 0.505029 seconds
 - Download time: 0.000844134 seconds
GPU GA*: Gold test failed!
 - Path length CPU: 580, GPU: 581
 - Path cost CPU: 728.528, GPU: 742.369
```

Boost Compute

Pure Template library, brings RAII, type-safety, common interface

```
// get the default compute device
compute::device gpu = compute::system::default device();
// create a compute context and command queue
compute::context ctx(gpu);
compute::command queue queue(ctx, gpu);
// generate random numbers on the host
std::vector<float> host vector(1000000);
std::generate(host vector.begin(), host vector.end(), rand);
// create vector on the device
compute::vector<float > device vector(1000000, ctx);
// copy data to the device
compute::copy(
    host vector.begin(), host vector.end(), device vector.begin(), queue
):
// sort data on the device
compute::sort(
    device vector.begin(), device vector.end(), queue
);
// copy data back to the host
compute::copy(
    device_vector.begin(), device_vector.end(), host_vector.begin(), queue
):
```

https://github.com/boostorg/compute

References

- [1] Andre Silva, Fernando Rocha, Artur Santos, Geber Ramalho, and Veronica Teichrieb. Gpu pathfinding optimization. In Games and Digital Entertainment (SBGAMES), 2011 Brazilian Symposium on, pages 158–163. IEEE, 2011.
- [2] Yichao Zhou and Jianyang Zeng. Massively parallel a* search on a gpu. In AAAI, pages 1248–1255, 2015.

In case you want to check it out:
https://github.com/Kruecke/ocl-astar