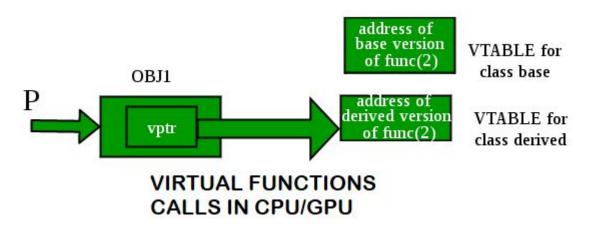
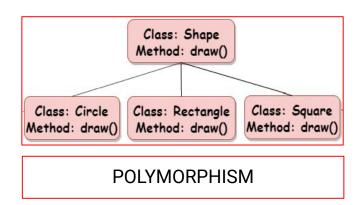
Judging a Type by Its Pointer: Optimizing GPU Virtual Functions

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Introducción





Problema del Paper







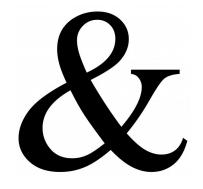


No Research in GPU's Applications

SharedOA Framework

Propuestas del Paper

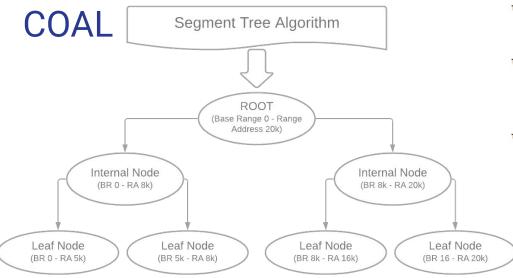






Hardware Support MMU Mod

Coordinated Object Allocation and function Lookup



Algorithm 1: Scan algorithm for the virtual range table

Function ObjectRangeLookup(objectAddr, funcIndex)

Complexity: O(log₂ (K))

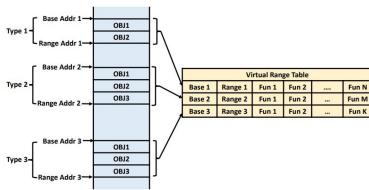
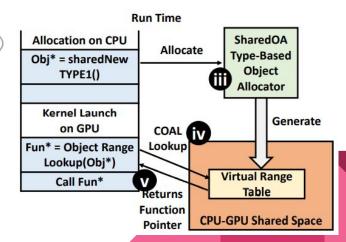


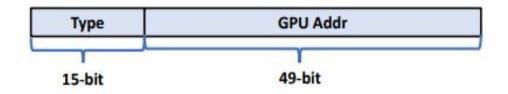
Figure 3: Type-based object allocator.



TypePointer



TypePointer Format



	<u> </u>	* ·	
1: <i>SHR</i>	Ra,	R_{obj} , #49	
2: <i>ADD</i>	Ra,	Ra, R _{vTablesStartAddr}	
3: <i>LDG</i>	Ra,	#vFuncOffset (Ra)	
4: CALL	Ra		

(b) Instructions implementing virtual function calls with Type-Pointer

Metodología



TypePointer

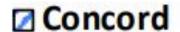


GPGPU-SIM

Hardware de Pruebas, con limitación de 4GB como HEAP size máximo

Workloads: Dynasoar, GraphChi-vE, Open Source Ray Tracer











Experimentación

Workload	Description	# Objects	# Types	vFuncs
	Dynasoar Workloads [22, 46]	X0.		
Traffic (TRAF)	A Nagel-Schreckenberg model traffic simulation to model streets, cars and traffic lights for traffic flows.	1573714	6	74
Game Of Life (GOL)	Of Life (GOL) Game of life is a cellular automaton formulated by John Horton Conway. This benchmark has two abstract classes Cells and Agent.		4	29
Structure (STUT)	Structure uses the Finite element method to simulate the fracture in a material. The benchmark models the material with springs and nodes.		4	40
Generation (GEN)	Generation is an extension of gol benchmark. The cells in Generation have more intermediate states which lead to more complicated scenarios.	1048576	4	33
	GraphChi-vE Workloads [35]	***	Ś.	
Breadth First Search (BFS)	BFS traverses graph nodes and updates a level field in a breadth-first manner. The GraphChi-vE BFS implementation defines an abstract class for edges, ChiEdge, and a concrete classEdge, which implements all the virtual functions of ChiEdge.	2254419	4	5
Connected Components (CC)	Connected Component is commonly used for image segmentation and cluster analysis, it employs an iterative node updates according to the labels of adjacent nodes.	2254419	4	6
Page Rank (PR)	Page rank is a classic algorithm to rank the pages of search engine results using iterative updates for each node.	2254419	4	3
	GraphChi-vEN Workloads [36]	***		
Breadth First Search (BFS)	The GraphChi-vEN BFS implementation also defines an abstract base class for vertex, ChiVertex, and a concrete class vertex, which implements ChiVertex's virtual functions.	2254419	4	15
Connected Components (CC)	nected Components (CC) GraphChi-vEN CC is similar to GraphChi-vE described above. However, GraphChi-vEN CC has both virtual edges and nodes.		4	15
Page Rank (PR)	GraphChi-vEN PR is similar to GraphChi-vE described above. However, GraphChi-vEN PR has both virtual edges and nodes.	2254419	4	10
	Open Source Ray Tracer [40]			
Raytracing (RAY)	RAY performs global rendering of of spheres and planes. The algorithm traces light rays through a scene, bouncing them off of objects, and back to the screen.	1000	3	3

Resultados

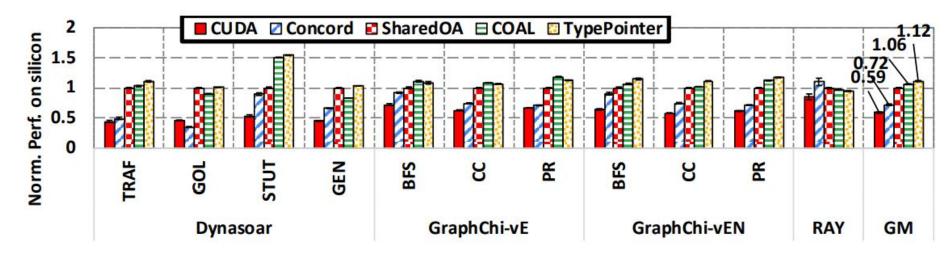


Figure 6: Performance, normalized to SharedOA on a silicon V100 GPU, averaged over 10 runs (error-bars=max and min).

Resultados

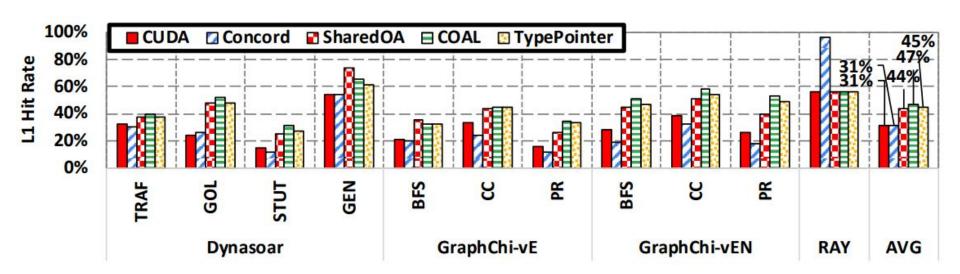


Figure 9: L1 cache hit rate on a silicon V100 GPU.

Conclusiones



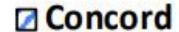


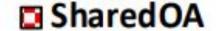


PERFORMANCE: 80%, 47% & 6% IMPROVEMENT PERFORMANCE: 90%, 56% & 12% **IMPROVEMENT**

GREAT WORK!









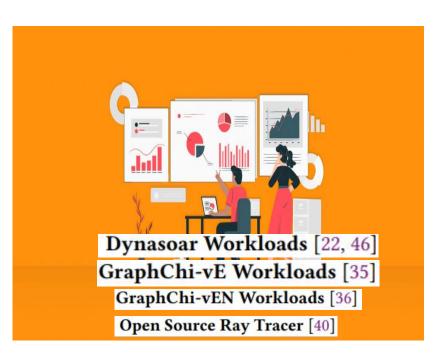


Descripción de las dependencias para las réplicas del trabajo

- **Hardware Dependencies**: Intel x86 machine with NVIDIA Volta architecture GPU with at least 16GB GPU memory. The experiments in this paper use V100 GPU with 32GB GPU memory. 1GB CPU memory is required to contain the evaluation repositories.
- **Software Dependencies**. Ubuntu 18.04 Linux is preferred to run the experiments. NVIDIA CUDA 10.1, 10.2 and 11.1 and GPU driver are required to compile and run the GPU workloads.

Experimentación

Utilización de los Workflow (Benchmarking):



Experimentos con ejemplos simples:

SharedOA, COAL and TypePointer

Software prerequisite

- Ubuntu 18.04.5 LTS Linux
- git
- Python 2.7
- CUDA 10.1: https://developer.nvidia.com/cuda-10.1-download-archive-base
- Transform script for COAL and TypePointer: https://github.com/bradmengchi/asplos_2021_ae

Experimentación con Ejemplos Simples (COAL)

```
#include "COAL.h"
#define NUM_OBJ 512
class S1 {
public:
 int var;
 host device S1() {}
 virtual __host__ _device__ void inc() = 0;
 virtual host___device__ void dec() = 0;
class S2 : public S1 {
public:
  __host___device__S2() {}
 _host __device _ void inc() { this->var += 2; }
  host device void dec() { this->var -= 2; }
```

```
21    _global__ void kernel(S1 **ptr) {
22    int tid = threadIdx.x + blockDim.x * blockIdx.x;
23    // this variable must be defined in every kerenl that uses COAL
24    void **vtable;
25    if (tid < NUM_OBJ) {
26        COAL_S1_inc(ptr[tid]);
27        ptr[tid]->inc();
28    }
29  }
```

```
my_obj_alloc.toDevice();
int blockSize = 256;
int numBlocks = (NUM_OBJ + blockSize - 1) / blockSize;
kernel<<<numBlocks, blockSize>>>(ptr);
cudaDeviceSynchronize();
```

Réplica del Experimento en COAL

Réplica con el Cluster de Google Colab:

Google Colab Experiment

```
Driver Version: 460.32.03
                                              CUDA Version: 11.2
               Persistence-M Bus-Id
                                       Disp.A | Volatile Uncorr. ECC
  Fan Temp Perf Pwr:Usage/Cap
                                  Memory-Usage
                                               GPU-Util Compute M.
                                                           MIG M.
 Tesla T4
                            00000000:00:04.0 Off
                                0MiB / 15109MiB
                                                          Default
  Processes:
  GPU GI CI
                    PID Type
                             Process name
                                                        GPU Memory
  No running processes found
nvcc: NVIDIA (R) Cuda compiler driver
Copyright (c) 2005-2020 NVIDIA Corporation
Built on Wed Jul 22 19:09:09 PDT 2020
Cuda compilation tools, release 11.0, V11.0.221
Build cuda 11.0 bu.TC445 37.28845127 0
```

```
vtable [252][0]:0x8
vtable [252][1]:0x10
Objects Creation Done
 Host Call Done
Device Call Done
Host Call Done
Device Call Done
ptr[0].var = 0
ptr[1].var = 0
 ptr[2].var = 0
 ptr[3].var = 0
ptr[505].var = 0
ptr[506].var = 0
ptr[507].var = 0
ptr[508].var = 0
ptr[509].var = 0
ptr[510].var = 0
ptr[511].var = 0
```

```
Type#0:
Type Size: 16 Number of Buckets : 1 Range Size: 1048576 Number of Objs : 512

Avg Types Size 16.000000
```

Experimentación con Ejemplos Simples (SharedOA)

```
#include "mem alloc.h"
#define NUM OBJ 512
class S1 {
public:
 int var;
  host device S1() {}
 virtual host device void inc() = 0;
 virtual host device void dec() = 0;
};
class S2 : public S1 {
public:
 host device 52() {}
  host device void inc() { this->var += 2; }
  _host__ device__ void dec() { this->var -= 2; }
};
```

```
__global__ void kernel(S1 **ptr) {
  int tid = threadIdx.x + blockDim.x * blockIdx.x;
  if (tid < NUM_OBJ)
    ptr[tid]->inc();
}
```

```
my_obj_alloc.toDevice();
int blockSize = 256;
int numBlocks = (NUM_OBJ + blockSize - 1) / blockSize;
kernel<<<numBlocks, blockSize>>>(ptr);
cudaDeviceSynchronize();
printf("Device Call Done\n");
```

Réplica del Experimento en SharedOA

Réplica en Google Colab:

```
vtable [252][0]:0x8
vtable [252][1]:0x10
Objects Creation Done
Host Call Done
Device Call Done
Host Call Done
Device Call Done
ptr[0].var = 0
ptr[1].var = 0
ptr[2].var = 0
ptr[3].var = 0
ptr[4].var = 0
ptr[5].var = 0
ptr[6].var = 0
ptr[7].var = 0
ptr[8].var = 0
ptr[9].var = 0
```

```
ptr[508].var = 0
ptr[509].var = 0
ptr[510].var = 0
ptr[511].var = 0
```

```
Type#0:
Type Size: 16 Number of Buckets: 1 Range Size: 1048576 Number of Objs: 512

Avg Types Size 16.000000
```

Experimentación con Ejemplos Simples (TP)

```
#include "mem_alloc_tp.h"
#include "TP.h"
#define NUM OBJ 512
class S1 {
public:
 int var;
 host device S1() {}
 virtual __host __device __void inc() = 0;
 virtual host device void dec() = 0;
};
class 52 : public 51 {
public:
 host device S2() {}
 _host __device void inc() { this->var += 2; }
  host device void dec() { this->var -= 2; }
```

```
_global__ void kernel(S1 **ptr) {
  int tid = threadIdx.x + blockDim.x * blockIdx.x;
  // this variable must be defined in every kerenl that uses COAL
  void **vtable;
  if (tid < NUM_OBJ) {
    S1 * obPtr = ptr[tid];
    TP_S1_inc(obPtr);
    CLEANPTR(obPtr,S1 *)->inc();
  }
}
```

```
my_obj_alloc.toDevice();
int blockSize = 256;
int numBlocks = (NUM_OBJ + blockSize - 1) / blockSize;
kernel<<<<numBlocks, blockSize>>>(ptr);
cudaDeviceSynchronize();
```

Réplica del Experimento en TP

Réplica en Google Colab:

```
vtable [252][0]:0x8
vtable [252][1]:0x10
total_count 1048576 , total 1048576 , type_size 16
Objects Creation Done
Type0: (nil)
Type1: 0x8
##############
gpuptr 0x7f78dc001138
Device Call Done
ptr[0].var = 2
ptr[1].var = 2
```

```
ptr[510].var = 2
ptr[511].var = 2
Type#0:
Type Size: 16 Number of Buckets : 1 Range Size: 1048576 Number of Objs : 512
Avg Types Size 16.000000
```

Link del Repositorio

Github

<u>Paper</u>