# Heterogeneous Parallel Programming COMP4901D

Data-Parallel Primitives:

Scatter and Gather

#### Overview

- Data-Parallel Primitives
  - Map, Prefix Scan, Scatter, Gather, Split, Sort
  - Others: Reduce, Filter, Search...
- GPU-Based Implementations
  - Gather and Scatter

# Processing a Large Number of Data Items

```
//sequential
  for (i = 0; i < N; i++)
         h C[i] = h A[i] + h_B[i];
//data-parallel
  global void VecAdd(int* A, int* B, int* C)
  int i = blockDim.x * blockIdx.x + threadIdx.x;
  C[i] = A[i] + B[i];
```

# Map and Prefix Scan

Primitive: Map

**Input**:  $R_{in}[1, ..., n]$ , a map function fcn.

**Output**:  $R_{out}[1,...,n]$ .

**Function**:  $R_{out}[i] = fcn(R_{in}[i])$ .

Primitive: Prefix Scan

**Input**:  $R_{in}[1, ..., n]$ , binary operator  $\bigoplus$ .

Output:  $R_{out}[1,...,n]$ . Function:  $R_{out}[i] = \bigoplus_{j \le i} R_{in}[j]$ .

#### Scatter and Gather

```
Primitive: Scatter
```

**Input**:  $R_{in}[1, ..., n], L[1, ..., n]$ .

**Output**:  $R_{out}[1, ..., n]$ .

**Function**:  $R_{out}[L[i]] = R_{in}[i], i=1, ..., n$ .

Primitive: Gather

**Input**:  $R_{in}[1, ..., n], L[1, ..., n].$ 

**Output**:  $R_{out}[1, ..., n]$ .

**Function**:  $R_{out}[i] = R_{in}[L[i]], i=1, ..., n$ .

# Split and Sort

```
Primitive: Split 

Input: R_{in}[1, ..., n], func(R_{in}[i]) \in [1,...,F], i=1, ..., n.

Output: R_{out}[1, ..., n].

Function: \{R_{out}[i], i=1,..., n\} = \{R_{in}[i], i=1, ..., n\} and func(R_{out}[i]) \leq func(R_{out}[j]), \forall i,j \in [1,...,n], i \leq j.
```

```
Primitive: Sort 

Input: R_{in}[1, ..., n]. 

Output: R_{out}[1, ..., n]. 

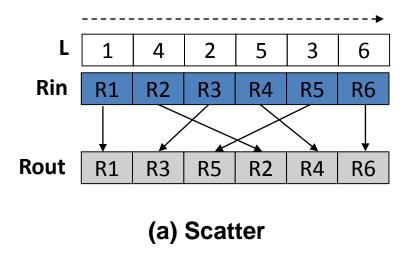
Function: \{R_{out}[i], i=1, ..., n\} = \{R_{in}[i], i=1, ..., n\} and R_{out}[i] \le R_{out}[j], \forall i, j \in [1,...,n] \text{ and } i \le j.
```

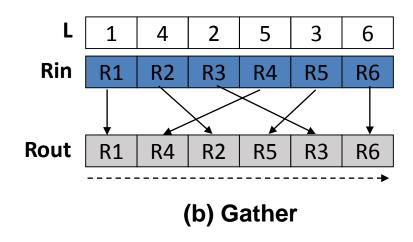
#### Scatter and Gather: Overview

- Widely supported
  - Parallel programming languages, e.g., MPI, NESL, ZPL.
  - Supercomputers, e.g., Cray MTA, Stanford Merrimac
  - Commodity co-processors (IBM Cell, GPUs)
- Irregular access patterns
  - Sparse matrix computations, hashing, searching, etc.
- Performance is memory bandwidth limited
  - Require high bandwidth architectures
  - HPC benchmarks (HPC Challenge, NAS PB, etc.)

#### **Access Patterns**

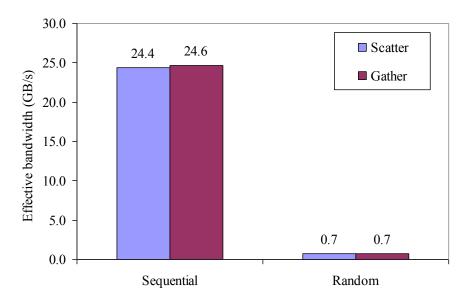
- Scatter: sequential reads and random writes.
- Gather: random reads and sequential writes.

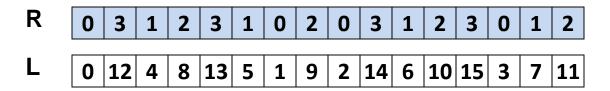




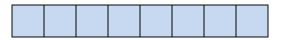
#### Scatter and Gather on the GPU

 Access pattern makes a 30X difference in performance.

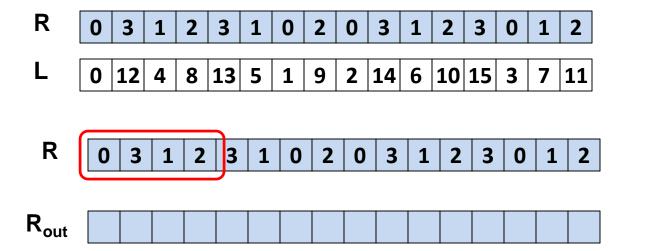




- 4 partitions
- 4 concurrent threads.

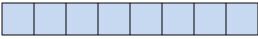


Cache



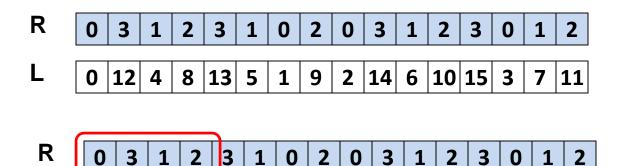
4 partitions

4 concurrent threads.



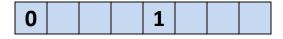
Cache

3



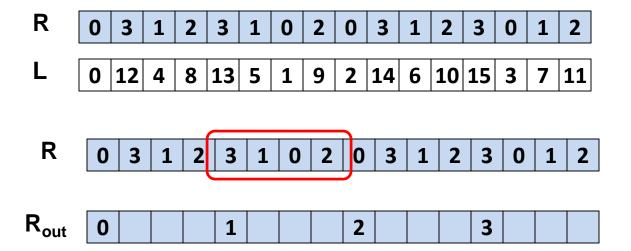
 $R_{\text{out}}$ 

4 partitions4 concurrent threads.



#### Cache

Cache Misses = 4 Cache Hits = 0



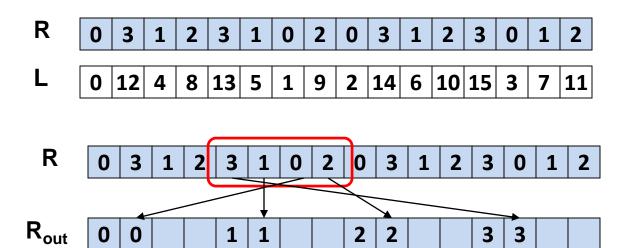
4 partitions

4 concurrent threads.

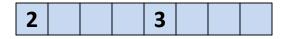


Cache

Cache Misses = 4 Cache Hits = 0

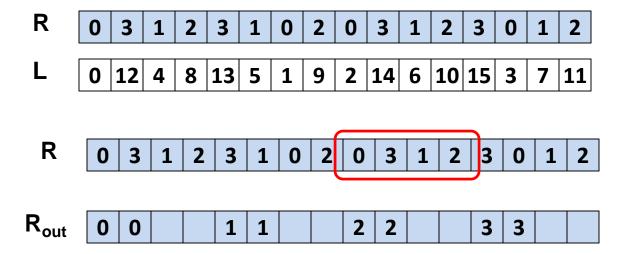


- 4 partitions
- 4 concurrent threads.

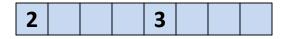


#### Cache

Cache Misses = 6 Cache Hits = 2

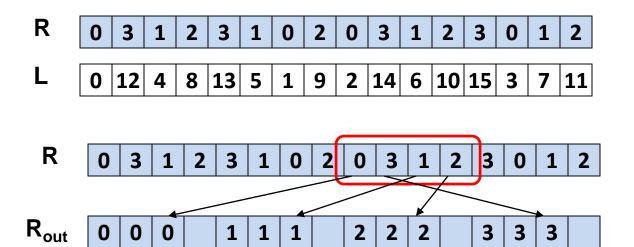


- 4 partitions
- 4 concurrent threads.

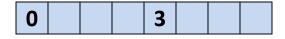


#### Cache

Cache Misses = 6 Cache Hits = 2

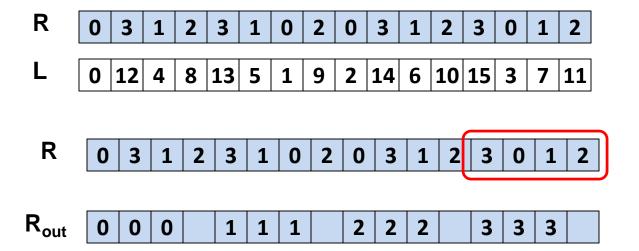


- 4 partitions
- 4 concurrent threads.



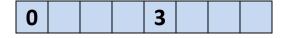
#### Cache

Cache Misses = 8 Cache Hits = 4



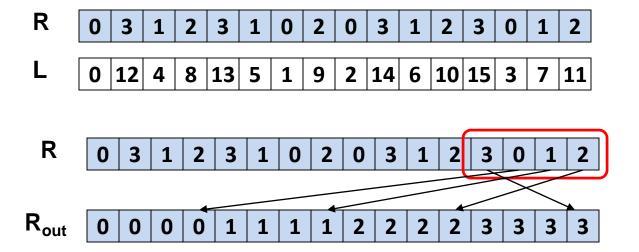
4 partitions

4 concurrent threads.



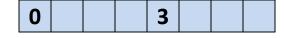
#### Cache

Cache Misses = 8 Cache Hits = 4



4 partitions

4 concurrent threads.



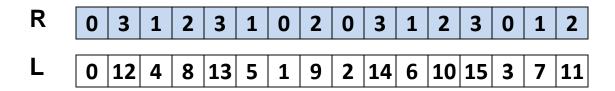
Cache

Cache Misses = 10 Cache Hits = 6

Cache miss rate = 62.5%Effective bandwidth =  $0.4 B_{seq}$ 

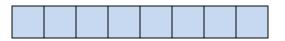
# Multi-pass Scheme

- The entire scatter is performed in multiple passes.
- Each pass writes to a small chunk

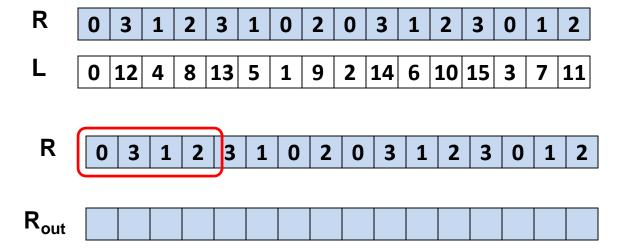


4 partitions

4 concurrent threads.

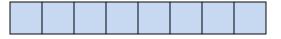


Cache

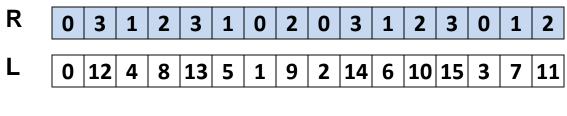


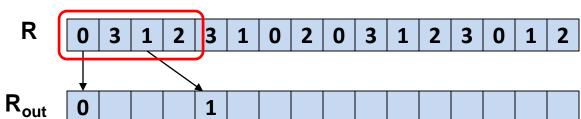
4 partitions

4 concurrent threads.

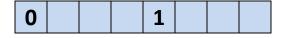


Cache



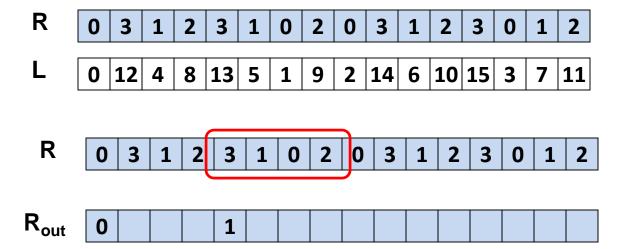


- 4 partitions
- 4 concurrent threads.



#### Cache

Cache Misses = 2 Cache Hits = 0



- 4 partitions
- 4 concurrent threads.

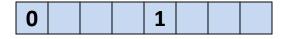


#### Cache

Cache Misses = 2 Cache Hits = 0

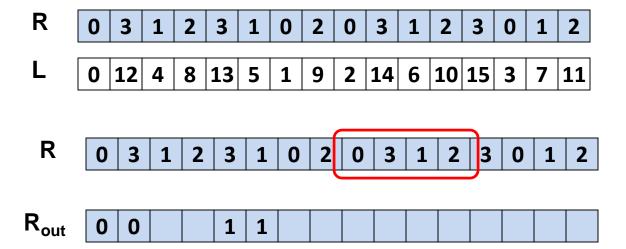


- 4 partitions
- 4 concurrent threads.

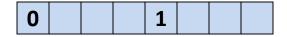


#### Cache

Cache Misses = 2 Cache Hits = 2

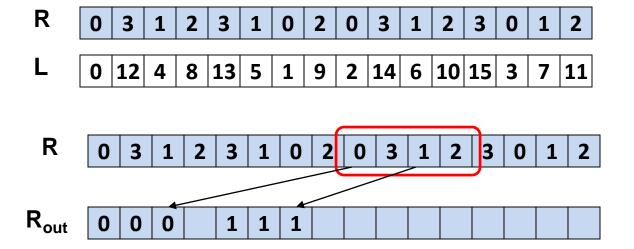


- 4 partitions
- 4 concurrent threads.

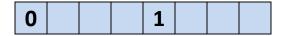


#### Cache

Cache Misses = 2 Cache Hits = 2

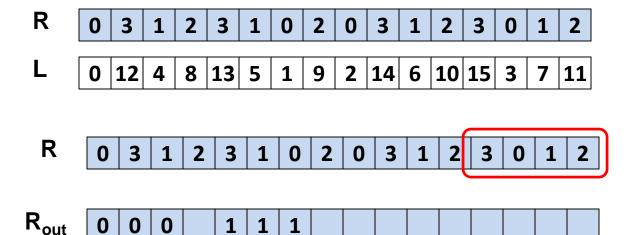


- 4 partitions
- 4 concurrent threads.



#### Cache

Cache Misses = 2 Cache Hits = 4

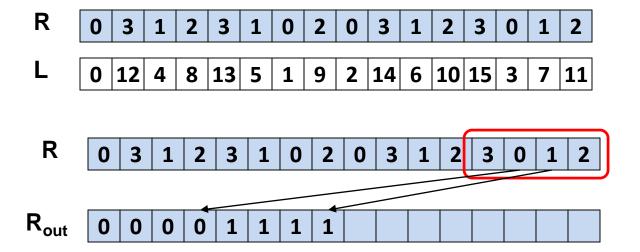


- 4 partitions
- 4 concurrent threads.

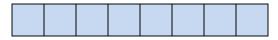


#### Cache

Cache Misses = 2 Cache Hits = 4

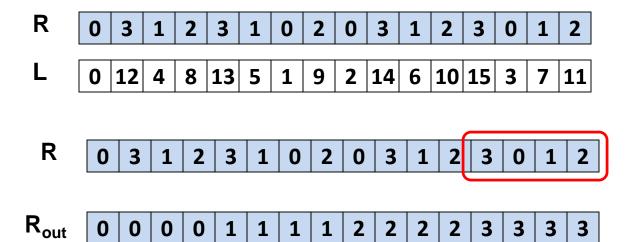


- 4 partitions
- 4 concurrent threads.



#### Cache

Cache Misses = 2 Cache Hits = 6



4 partitions

4 concurrent threads.



Cache

Cache Misses = 4 Cache Hits = 12

Cache miss rate = 25% Effective bandwidth = B<sub>seq</sub>

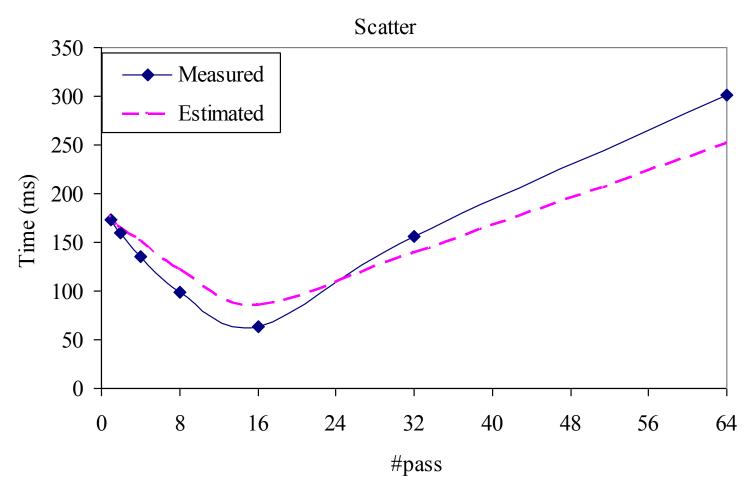
#### Cost Model

- Estimate the performance of different access patterns
  - Sequential bandwidth
  - Random bandwidth
- Estimate the total cost of sequential access and random access in the multi-pass scheme.

$$T_{\text{scatter}} = (|R| + |L|) * \text{npasses/B}_{\text{seq}} + |R|/B_{\text{rand}}$$

Determine the optimal number of passes.

# Performance Results -- Multi-pass Scatter



The optimal number of passes is 16.

# Applications and Analysis

#### Applications

Radix sort, hash search, and sparse-matrix vector multiplication

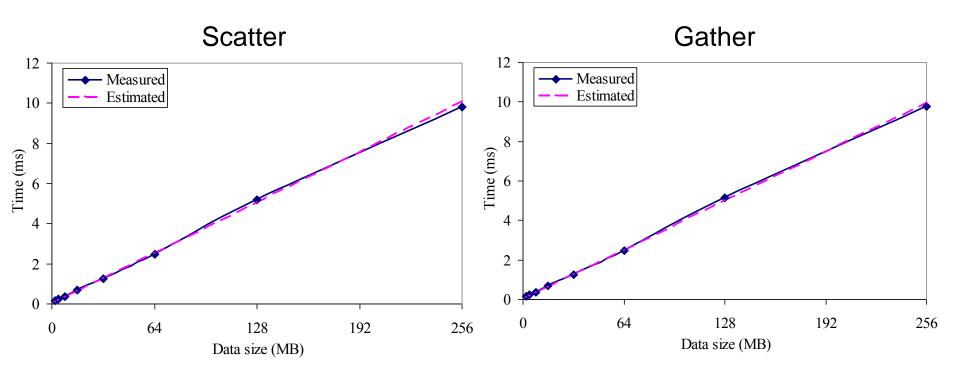
#### Platforms

- CPUs: Intel Quad, or two AMD dual-core processors.
- GPU: Nvidia 8800 GTX.

#### Overall results

- The cost model has an accuracy of over 85%.
- The multipass scheme improves the application 10%~50%.
- The GPU-based algorithm outperforms the CPU-based algorithm by 2-7X.

# Sequential Scatter/Gather



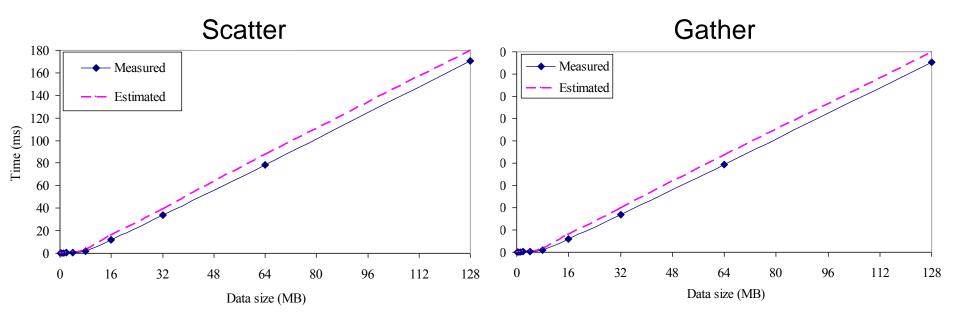
Accuracy of the cost model on the sequential gather and the scatter: 87%

Accuracy = 
$$1 - \frac{|Measuremen t - Estimation|}{Measuremen t}$$

Qiong Luo

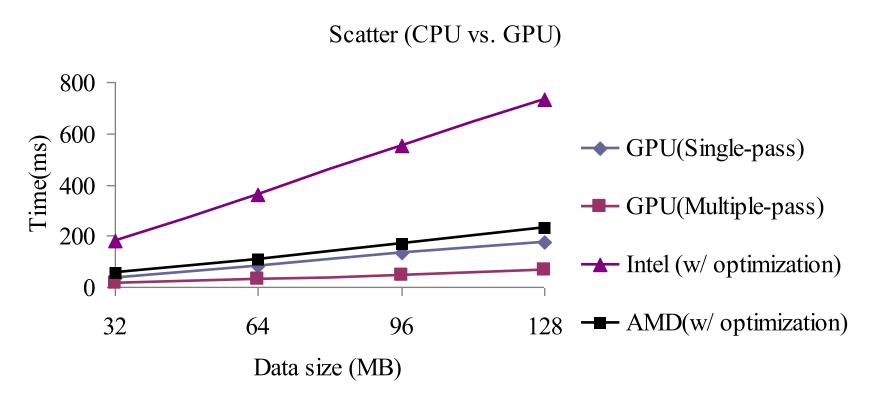
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# Random Scatter/Gather



Accuracy of the cost model on the random gather and the scatter: 90%.

#### Performance Impact of Multi-Pass Scatter



- (1) The speedup is 7-13X and 2-4X on Intel and AMD, respectively.
- (2) The multi-pass scheme improves the GPU-based scatter by 2-4X.

## Summary

- Data-parallel primitives are an efficient way of utilizing GPU's parallelism.
- Scatter and gather are memory-bound and can be optimized through multi-pass schemes.

#### References:

Bingsheng He, Naga K. Govindaraju, Qiong Luo, and Burton Smith. Efficient Gather and Scatter Operations on Graphics Processors. ACM/IEEE SuperComputing (SC), Nov 2007.

http://www.cse.ust.hk/gpuqp