# GraphReduce: Processing Large-Scale Graphs on Accelerated-Based Systems

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### **Graphs are Ubiquitous**



- High Volume: Billions of edges and vertices carrying rich metadata
- High Velocity: 100s of billions photos, posts, tweets etc per month
- Fast graph analytics on large graphs

### **GAS Programming Model**

### Motivation

- Why use GPUs ? GPU-based frameworks are orders of magnitude faster
- Previous GPU-based graph processing doesn't handle datasets that doesn't fit in GPU memory
  - Yahoo-web graph with 1.4 billion vertices requires 6.6 GB memory just to store its vertex values.
- Several challenges in large-scale graph processing
  - How to partition the graph ?
  - How and when to move the partitions between host and GPU ?
  - How to best extract multi-level parallelism in GPUs ?

### **Hybrid Programming Model**

Graphs	X-Stream (ms)	CuSha(ms)	Speedup
ak2010	215.155	7.75	28x
belgium_osm	2695.88	791.299	3x
coAuthorsDBLP	1275	11.553	110x
delaunay_n13	80.89	5.184	16x
kron_g500-logn20	46550.7	119.824	389x
webbase-1M	3909.12	13.515	290x



 Scatter phase: each vertex updates the state of every outgoing edge.

### vertex scatter (vertex v) : edge scatter (edge e) : send updates over outgoing edges of v send update over e vertex gather (vertex v) : update gather (update u) : apply updates from inbound edges of v apply update u to u destination while not done : while not done : for all vertices v that need to scatter updates for all edges e vertex scatter (v) edge\_scatter (e) for all updates u for all vertices v that have updates update gather (u) vertex gather (v)

### Vertex-centric GAS

Edge-centric GAS

- Existing systems choose either vertex- or edge-centric GAS programming model for graph execution
- Different processing phases have different types of parallelism and memory access characteristics
- GraphReduce adopts a hybrid model with a combination of both vertex- and edge-centric model

## Optimizations



### **GraphReduce Architecture**





- User defined functions: gatherMap(), gatherReduce(), apply() and scatter()
- User defined graph data types : VertexDataType and EdgeDataType

### Results



Benefits of GraphReduce optimizations over memcpy time

- HYPER-Q
- Asynchronous execution and Spray (deep-copy) operation
- Dynamic frontier management
- Dynamic phase fusion and elimination

### Conclusions

- **GraphReduce** develops a high performance graph processing framework for input datasets that may or may not fit in GPU memory
- Adopts a hybrid model of a combination of both edge- and vertexcentric implementation of GAS programming model
- Leverages CUDA streams and hardware supports like hyper-Qs to stream data in and out of GPU for high performance
- Optimizations like dynamic phase fusion/elimination and frontier management further reduces data transfer time
- Outperforms CPU-based out-of-core graph processing frameworks across a variety of real data sets achieving up to 79x speedup



