GARAPH EFFICIENT GPU-ACCELERATED GRAPH PROCESSING ON A SINGLE MACHINE WITH BALANCED

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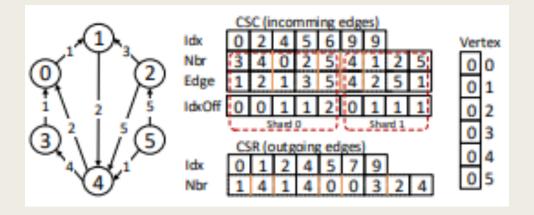
REPLICATION

Goals of Garaph

- Use all resources of machines
 - Large Memory
 - Fast Secondary Storage
 - CPU
 - GPU
- Prior issues
 - Skewed degree distribution
 - Write contention
 - Work imbalance

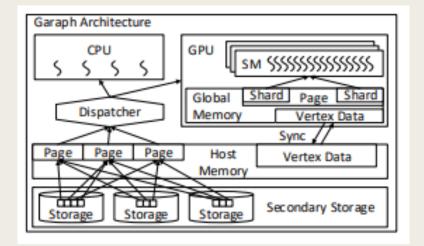
Graph Representation

- Shards
 - Disjoint set of vertices along with the incoming edges
 - Increasing order of destination
 - Sized to fit in the GPUs shared memory (about 12k vertices)
 - Replicated
- Pages
 - A set of shard for efficient movement of data
- CSC for incoming edges
 - pull
- CSR for outgoing edges
 - Poor behavior on GPUs
 - Notify pull
 - Only neighbors of the active set pull

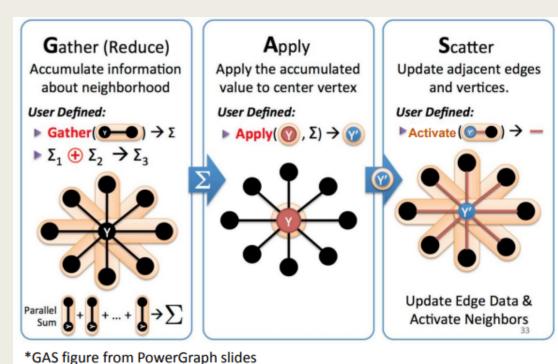


System Architecture

- Dispatcher
 - Loading graph
 - Distributing computation
 - Distribute data onto multiple SSDs
- GPU/CPU computation kernel
 - All blocks processed in parallel
 - Only pull on GPU
 - Both pull and notify pull on CPU
 - Can run either synchronously or asynchronously
 - When asynchronous updates are immediately visible
- Fault Tolerance
 - Periodically writes state to secondary storage



Programming API's



```
\begin{array}{l} \textbf{interface} \quad \texttt{GASVertexProgram}\left(\textbf{u}\right) \\ \textbf{gather}\left(D_{u}, D_{(u,v)}, D_{v}\right) \rightarrow \quad \texttt{Accum} \\ \textbf{sum}\left(\texttt{Accum} \quad \texttt{left}, \texttt{Accum} \quad \texttt{right}\right) \rightarrow \quad \texttt{Accum} \\ \textbf{apply}\left(D_{u}, \texttt{Accum}\right) \rightarrow D_{u}^{new} \\ \textbf{activate}\left(D_{u}^{new}, D_{u}\right) \rightarrow \quad A[u] \end{array}
```

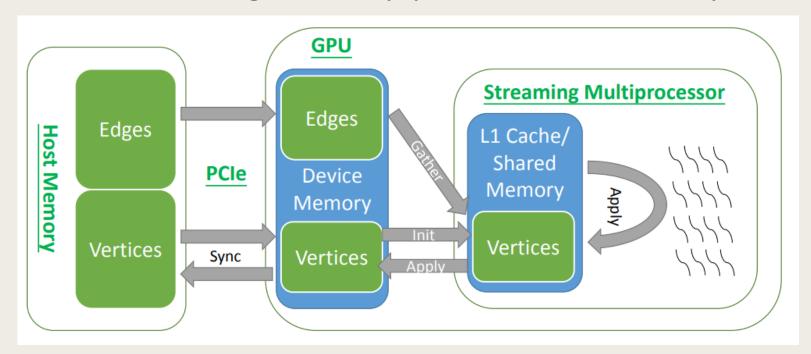
```
template < typename T>
    __host__ __device__ void unifiedAdd(T *addr, T val) {
#ifdef __CUDA_ARCH__ // For GPU, atomic operation
        atomicAdd(addr, val);
#else // For CPU, non-atomic operation
        *addr += val;
#endif
}
```

GPU-Based Graph Processing

- Graph Processing Engine
- Replication-Based Gather

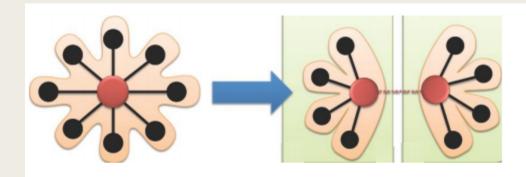
Graph Processing Engine

- Vertices stored in GPU global memory
- Each SM has a local copy of the vertices of the shard
 - Gather by reading from global memory updating the local copy
 - Then written back to global memory on GPU
 - After round GPU global memory synchronized with host memory

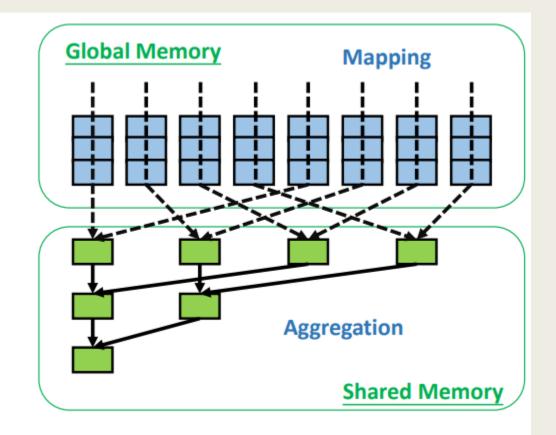


Replication-Based Gather

- To avoid write contention
 - Within a shard lots of edges going to the same node
 - Made worse by natural graphs power law distribution
 - Replicate the node data and sum up partial values then accumulate

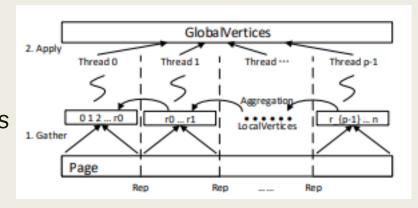


- Customized replication
 - O(N) -> O(log N), $N \le 32$
 - Modeling: balance profits and costs
 - $R_i = 2^{\min\left\{\left\lceil \log \frac{|E_i|}{|V_i|} 0.5\right\rceil, 5\right\}}.$



CPU-Based Graph Processing

- Processing with Edge Partitions
 - Edges or split up equally into different partitions
 - Vertexes split are duplicated
 - Later aggregated to obtain value
- Dual-Mode Processing Engine



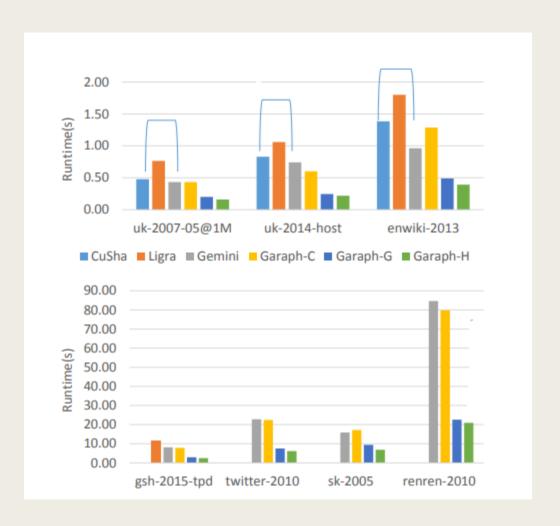
Evaluation

- Comparison with Other Systems
- Customized Replication
- Dual Modes of the CPU Kernel
- Hybrid CPU-GPU Scheduling

Graph	V	E	Max in-deg	Avg deg	Size edgelist
uk-2007@1M	1M	41M	0.4M	41	0.6GB
uk-2014-host	4.8M	51M	0.7M	11	0.8GB
enwiki-2013	4.2M	0.1B	0.4M	24	1.7GB
gsh-2015-tpd	31M	0.6B	2.2M	20	10GB
twitter-2010	42M	1.5B	0.8M	35	27GB
sk-2005	51M	1.9B	8.6M	39	35GB
renren-2010	58M	2.8B	0.3M	48	44GB
uk-union	134M	5.5B	6.4M	41	0.1TB
gsh-2015	988M	34B	59M	34	0.7TB

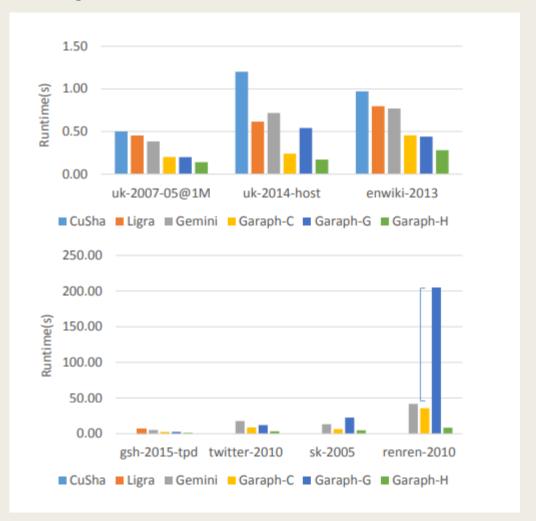
Comparison with Other Systems

■ 10 iterations Pagerank



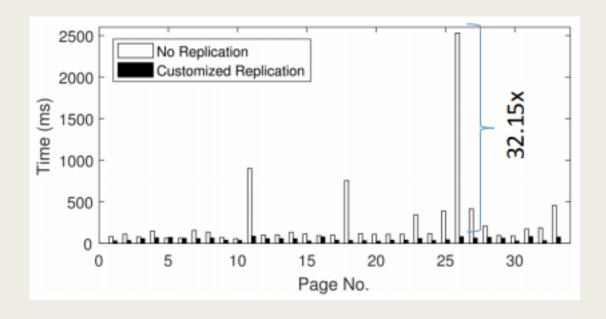
Comparison with Other Systems

- Connected components
- Until convergence
- GPU can be much slower



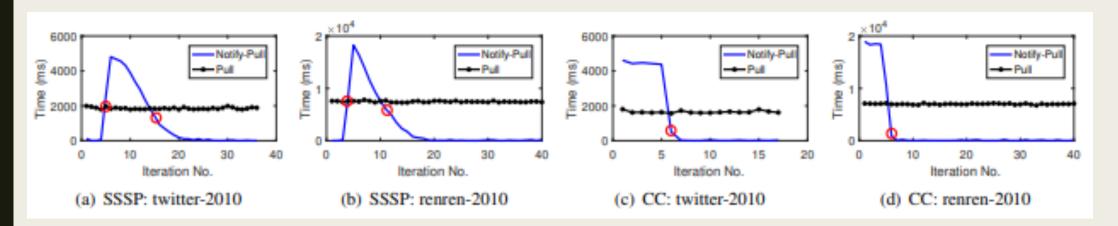
Customized Replication

Helps some pages dramatically



Dual Modes of the CPU Kernel

On some iterations notify pull is much better



Hybrid CPU-GPU Scheduling

