

High-Performance
Graph Processing:
Locality,
Vectorization and
Reduced Precision

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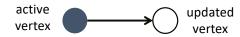
1 December 2020

### **GRAPH ALGORITHMS**

Iteratively calculate a property of each vertex in a graph

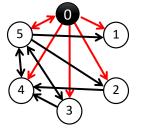
E.g.: PageRank values, label of connected components, embeddings in graph convolutional neural networks, etc.

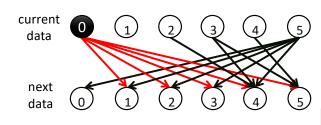




Driven by frontier: set of active vertices

Many updates in parallel, with conflicts







### **Programming Interface**

size(U: frontier): N

returns |U|

EdgeMap(G: graph,

U : frontier,

 $F: (vertex \times vertex) \rightarrow bool,$ 

C: vertex  $\rightarrow$  bool): frontier

VertexMap(U: frontier,

 $F : vertex \rightarrow bool) : frontier$ 

Ligra [Shun PPoPP'13] Assume graph G=(V,E)

EdgeMap applies an operation F to each edge  $(u,v) \in E$  where  $u \in U$  and C(v) = true. It returns a frontier that contains all v where any call to F(u,v) returned true

**VertexMap** applies an operation F to each vertex  $v \in U$  and returns a frontier that contains v iff  $v \in U$  and F(v) = true

In both cases, F may have side effects, e.g., updating properties for the vertices



# **EXAMPLE: CONNECTED COMPONENTS**



Question: How to implement vertexmap and edgemap efficiently?

# Part I: Non-Uniform Memory Architectures

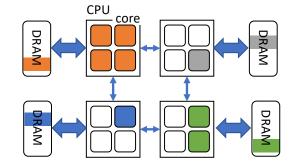


### **GOAL**

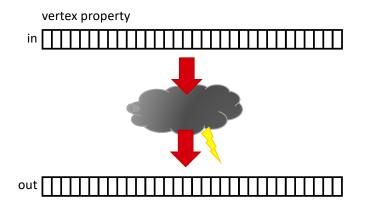
How to map graph analytics over immutable graphs onto a NUMA architecture while minimising execution time?



Remote access has higher latency, lower bandwidth than local access Stores are more affected than loads







# EDGEMAP, VERTEXMAP AND NUMA-AWARENESS

**Goal**: map code and data to NUMA nodes One type of arrays

• Properties (per vertex)

Two types of loops

- Loops over edges
- Loops over vertices

Two types of iteration

- Sparse frontier
- Dense frontier



### **RECAP: DATA RACES**

A pair of load and store instructions, at least one of which is a store, that access the same memory location

In a concurrent program with data races, the outcome of the program may differ depending on the relative execution speed of threads

### Typical solutions:

- mutual exclusion
- atomic memory operations
- owner-computes

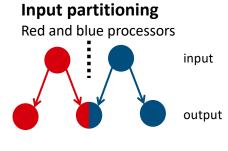


### **OWNER-COMPUTES**

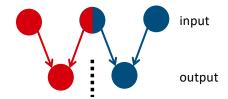
Decomposition based on partitioning input/output data is referred to as the owner computes rule

Each partition performs all the computations involving data that it owns

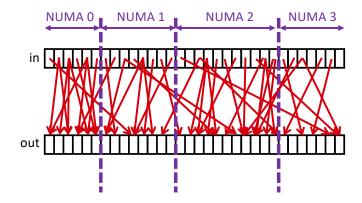
- Input data decomposition: A task performs all the computations that can be done using these input data
- Output data decomposition: A task computes all the results in the partition assigned to it



### **Output partitioning**







## NUMA-AWARE LAYOUT FOR EDGEMAP

#### Goal

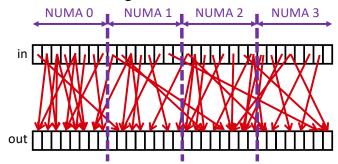
Determine cuts of { code, data } such that performance is maximised

#### How?

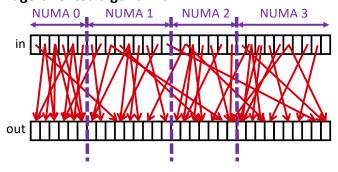
Partition graph such that each partition (NUMA node) has an equal:

- 1. #edges, #cuts [PowerGraph OSDI'12] ... breaks locality
- #sources [X-stream SOSP'13]... remote updates
- 3. #edges [Polymer PPoPP '15]
- 4. ( $\alpha$  #destinations + #edges) [Gemini OSDI'16]

### **Vertex-oriented algorithms**



#### **Edge-oriented algorithms**



## NUMA-AWARE LAYOUT FOR EDGEMAP

It depends! [GraphGrind ICS'17]

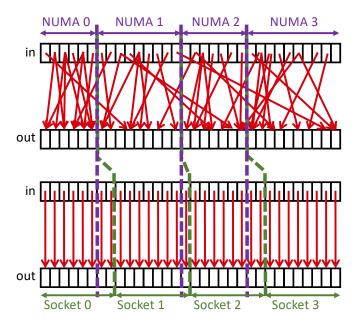
### "Vertex-oriented" algorithms

- Best performance with equal #destinations
- Frontier density mostly below 50%
- BFS, Betweenness Centrality, Bellman-Ford

### "Edge-oriented" algorithms

- Best performance with equal #edges
- Frontier density mostly close to 100%
- PageRank, SpMV, Belief Prop., PageRankDelta

### **Edge-oriented algorithms**



## NUMA-AWARE LAYOUT FOR VERTEXMAP

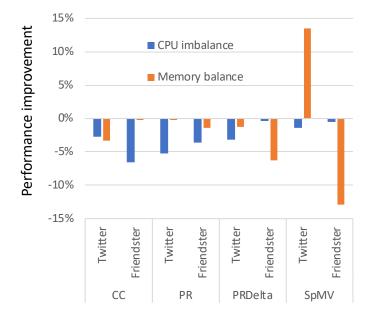
### "Vertex-oriented" algorithms

Trivial

### "Edge-oriented" algorithms

- Need to choose between balancing compute and minimising traffic across NUMA nodes
- Better to balance compute and incur additional inter-node traffic [GraphGrind ICS'17]
- Consequently, data is partitioned the same way as compute in edgemap, but differently in vertexmap

#### 4-socket 2.6GHz Intel Xeon E7-4860 v2, 48 threads, 256 GiB

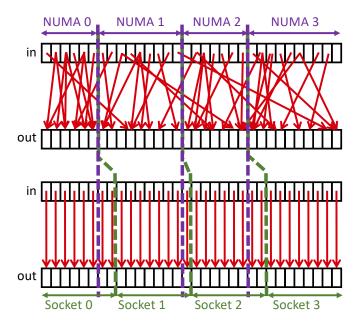


J. Sun, H. Vandierendonck and D. S. Nikolopoulos, "GraphGrind: addressing load imbalance of graph partitioning", ICS'17

#### **NUMA-AWARENESS CHOICES**

- Baseline is CPU balance and memory imbalance
  - Implies remote accesses during vertex map
- CPU imbalance
  - No remote accesses during vertex map
- Memory balance
  - No remote accesses during vertex map
  - Many remote accesses during edge map

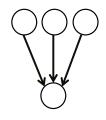
### **Edge-oriented algorithms**

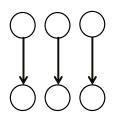


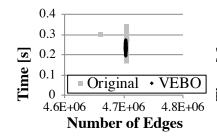
# CAN WE MEET BOTH REQUIREMENTS?

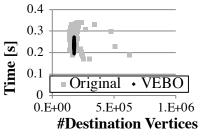
Have our cake and eat it too!

Revisiting edge balance: Two partitions with 3 edges Which partition is processed faster?









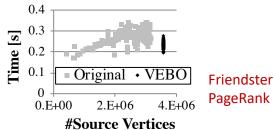
### **LOAD BALANCE**

Execution time/partition highly dependent on the degree of vertices

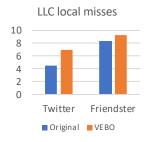
Reorder vertices

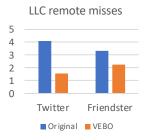
- in order of decreasing in-degree
- using list scheduling

VEBO: Vertex and Edge Balanced Partitioning

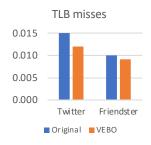


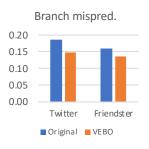
### **VEBO BENEFITS**











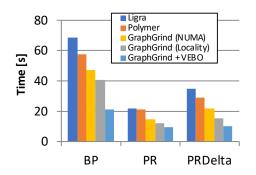
All metrics per thousand instructions

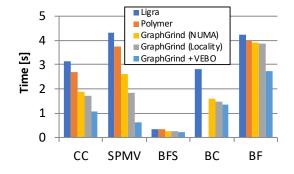
Partitions are processed faster as a side-effect of reordering

Remote cache misses are traded for local misses









### **PERFORMANCE**

Comparing Ligra, Polymer (NUMA-aware), and 3 versions of GraphGrind

Twitter graph

4-socket 2.6GHz Intel Xeon E7-4860 v2, 48 threads, 256 GiB

Similar results hold for other graphs

VEBO relabels vertex IDs to achieve load balance

