Cost Efficient Large-Scale Graph Analytics

Dr. Joseph Schneible

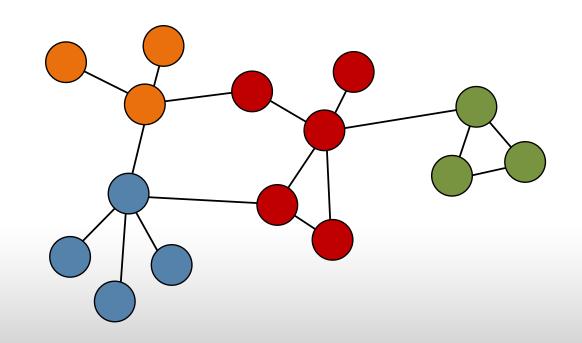




Applications of Graph Analysis



- Social Networks
- WWW
- Medicine
- Natural Language
- Cybersecurity
- Homeland Security
- Local Government





OUTLINE:

- Graph Analysis
- System Design
- Performance





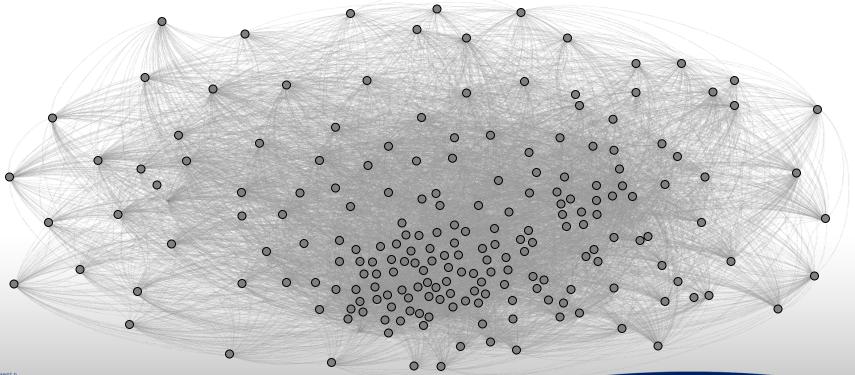
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How to bring meaning to this?







Example Algorithms

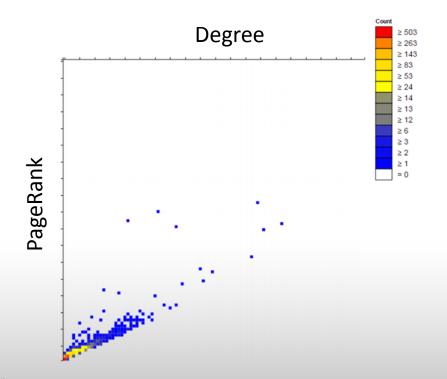


PageRank	Find Influential Nodes Within a Network
Community Detection	Find Dense Sub-graphs
Belief Propagation	Perform Inference on a Graph



PageRank





- PageRank trends linearly with degree
- Anomalous nodes are above this trend line
- Used to find mastermind of 9/11 attacks
- Can be applied to biological networks, etc



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Goals







Time Efficiency



Customizability



Meaning



System Approach



 All of these are affected by design choices:

Commodity Hardware

Memory Efficiency

GPU Utilization

I/O Efficiency Graph Construction



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Challenges



Parallelization	Memory Limitations	Irregular Graph Structure
 Edges or Vertices? 	GB of RAM and vRAMTB Graphs	 Many Nodes with Few Connections Few Nodes with Many Connections



Parallelization Strategies

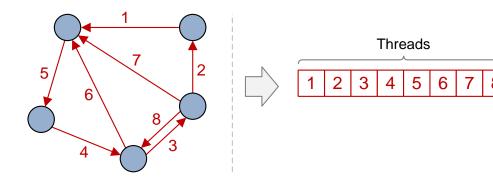


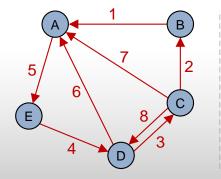
Edge-wise Distribution:

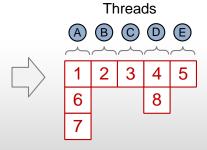
- One Operation per Edge
- Memory for Temporary Data Structures
- Even Load Balance

Vertex-wise Distribution:

- Multiple Operations per Vertex
- Uneven Load Balance





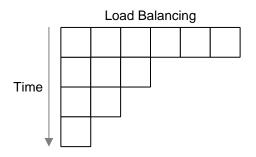




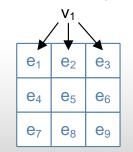
Load Balancing Graph Analysis on the GPU

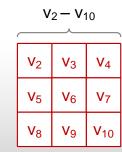


- High degree vertices will dominate computation time
- Created multiple kernels
- Threshold between high and low degree



Multiple Kernels



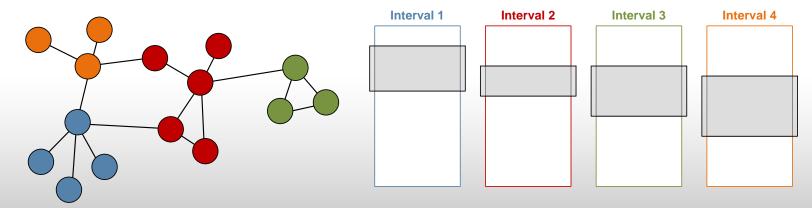




Out-of-Core Graph Processing



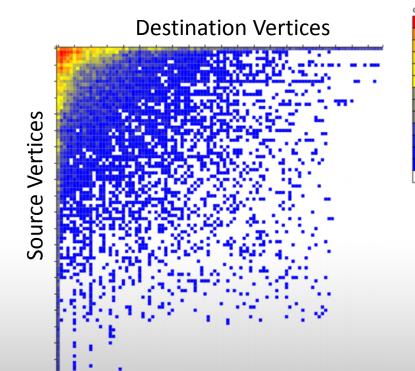
- Divide graph into intervals (sets) of vertices
- Gather associated in-edges into a shard
- Order edges in shards such that out-edges are located together in windows





Compression





Power-law distribution is common in natural graphs

≥ 72

≥ 50

≥ 39 ≥ 34 ≥ 31

≥ 29 ≥ 17 ≥ 16

≥ 15 ≥ 8

≥ 1

 Compression scheme exploits distribution





Task Analysis



Use:

- Graph Meta-data
- Performance Models
- Micro-benchmarks

■ To:

- Divide work between CPU and GPU
- Divide work between kernels





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PageRank Performance



PageRank:

5 Iterations

LiveJournal Graph:

■ Vertices: 4.6 Million

■ Edges: 77.4 Million



FUNL Desktop System

■ 9.5 Seconds



Spark Cluster

110.4 Seconds

FUNL Desktop System:

GPU — GeForce GTX TITAN, 2688 CUDA Cores, 928MHz, 6GB vRAM

CPU — Core i7, Quad Core, 3.40GHz

RAM — 16GB (4x4GB), 1333MHz

Storage — HDD, 180MB/s

Spark Cluster:

System: AWS EC2 m1.large

Number of Nodes: 10

Network: Moderate Performance



Belief Propagation Performance

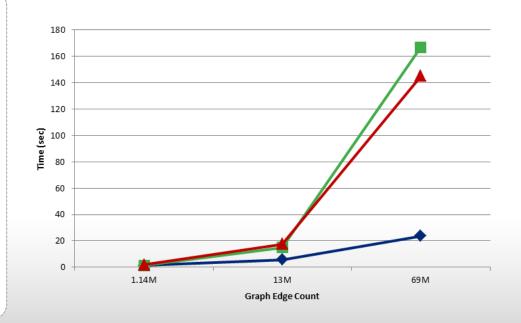


FUNL/Quad Core CPU:

- GPU GeForce GTX TITAN, 2688 CUDA Cores, 928MHz, 6GB vRAM
- CPU Core i7, Quad Core, 3.40GHz
- RAM 16GB (4x4GB), 1333MHz
- Storage HDD, 180MB/s

16 Core Server:

- CPU Xeon E5-2690, 16 Cores, 2.9GHz
- RAM 64GB, 1600MHz
- Storage HDDx6, RAID0, 690MB/s



- https://github.com/GraphChi



Additional Slides





Advantages



Unique Features:

- Large scale Graph Analysis on the GPU
- Task Analysis for Efficient Parallel Processing (on CPU's and GPU's)
- UI and Interactive Visualization to bring Meaning to Big Data

Benefits:

- Big Data Graph Analysis on a Budget
- Customizability
- Ease of Use (You don't have to be a data scientist)
- Reduction in Infrastructure and Energy Needs





Big Data Appliance for Graph Analytics



Gain insight by discovering unknown relationships in big data.

Graph analytics solution that supports pattern discovery and inferencing on large scale data sets.

Achieve a competitive advantage without a large budget.

Purpose-built to solve big data graph problems with commodity hardware.

Ease adoption with a small footprint solution and customizability.

Data-center-friendly appliance with a suite of graph algorithms and flexibility to add custom solutions.



Point of Contact



Joe Schneible

Enterprise Software Solutions Engineering Group Manager

Email: jschneible@technicacorp.com

LinkedIn: linkedin.com/in/jschneible



Technica Corporation 22970 Indian Creek Dr., Suite 500 Dulles, VA 20166

703.662.2000

technicacorp.com

