Monarch: Gaining Command on Geo-Distributed Graph Analytics

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Graph Analytics Popular















Graph Analytics Popular



GIRAPH GraphX







Social Networks



Cellular Network Analytics



Financial Network Analytics \$ BANK SYNTHETIC ACCOUNT PERSON 1 UNSECURED CREDIT LOAN CARD SYNTHETIC PERSON 2 _____ ACCOUNT HOLDER 1 -----PHONE SSN 1 NUMBER PHONE SSN 2 NUMBER ACCOUNT ADDRESS ACCOUNT HOLDER 2 HOLDER 3 \$ CREDIT UNSECURED BANK BANK CARD ACCOUNT ACCOUNT LOAN

Image courtesy: Neo4J





Generate data in a *geo-distributed* fashion





Generate data in a *geo-distributed* fashion

Can benefit from *timely* analysis

How do we perform efficient *geo-distributed* graph analytics?

Geo-Distributed Analytics (GDA)



Slide courtesy: Clarinet authors

Geo-Distributed Analytics (GDA)



Geo-Distributed Analytics on Graphs Can we use the same idea on graphs?

- GDA focuses on simple task placement/queries
 - Graph analytics iterative in nature
- Flexibility over data placement and join sites
 - Graph partitioning difficult
- Estimating intermediate data
 - Difficult in graph algorithms

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Key: Optimizing iterative graph-parallel processing



Gather: Accumulate information from neighborhood



Gather: Accumulate information from neighborhood Apply: Apply the accumulated value



Gather: Accumulate information from neighborhood Apply: Apply the accumulated value

Scatter: Update adjacent edges & vertices with new value













Execution Model





Execution Model

WAN Awareness

Graph Sparsification

- Sparsification extensively studied in graph theory
 - Idea: approximate the graph using a sparse, much smaller graph
 - Drop edges/vertices



- Sparsify without accuracy loss
 - Only worry about reducing cross-DC entities
 - Leverage graph-parallel model and algorithm properties





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Which graph algorithms can use the iGAS model? How much state needs to be kept at the entities for accuracy?









Evaluation of Potential

- 16 node Apache Spark cluster across 4 regions
- Modified GraphX to incorporate the proposed model



Other Open Questions

- Convergence properties due to our modified execution model
- Better execution models at bootstrap stage
 - How would the global sync work?
- Multi-tenancy
 - Would it provide opportunities to leverage existing GDA techniques?
- Graph updates
 - What is an incremental model in this case?

Conclusion

- Several emerging applications produce graph data in a geo-distributed fashion
 - Can benefit from geo-distributed graph analytics.
- Our proposal Monarch:
 - Early attempt at bringing geo-distributed analytics to graph processing.
 - Initial results are encouraging.

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