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Graph Analytics



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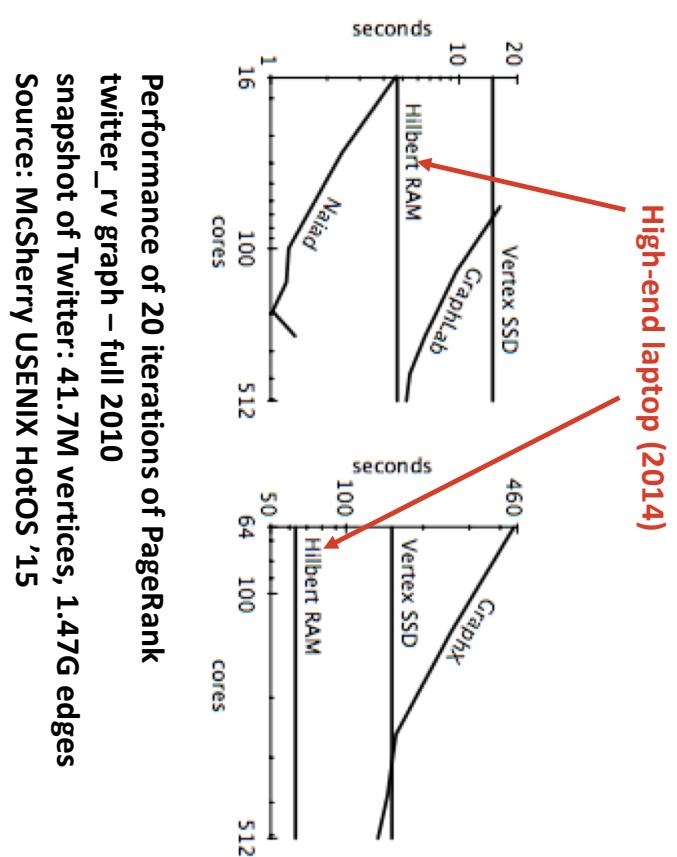
Graph Analytics and Cyber Security

- Insight from graph-structured data
- Representing interactions between entities as graphs
- Centrality metrics
- Anomaly detection
 - OddBall [PAKDD'10], Recursive structural features [KDD'11]
 - Outlier detection [KDD'10], PICs [SDM'12]
 - YAGADA [CIKM'11]
- Subgraph matching and enumeration [VLDB'12]
- Extrapolating labeled information to other vertices [*Pimplikar et al, CIKM'14*]



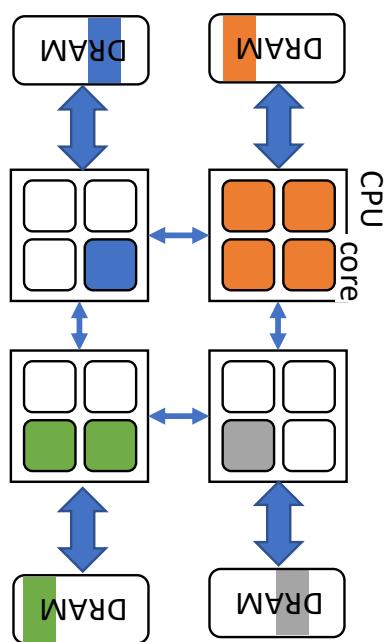
High-Performance Graph Analytics

- Timeliness: high-performance analytics
- Open source tools facilitate adoption:
 - Giraph, GraphLab/PowerGraph, Spark+GraphX
- Tools focus on scalability, not performance
- Scale-out; small clusters; data centers
 - McSherry: Scalability! But at what COST?
 - Compare against laptop!



The GraphGrind Framework: Fast Graph Analytics on Shared-Memory Systems

- The issue with scale-out systems
 - Workloads characterised by frequent communication and synchronisation
 - Can fall-back to processing from disk; likely competitive to distributed memory processing
- The alternative: a scaled-up server
 - Plenty of working memory possible; up to 16TB in shared memory configuration
 - Future: die stacking (e.g., hybrid memory cube), non-volatile memory (e.g., PCM, ReRAM)
 - But: must be a multi-socket (NUMA, non-uniform memory architecture) setup



GraphGrind

- Challenges:

- Balance threads and data over NUMA nodes
- Load balancing depends on graph algorithm, varies throughout computation
 - Tension between balancing #vertices vs #edges
 - Difficulty: memory locality and memory access latency

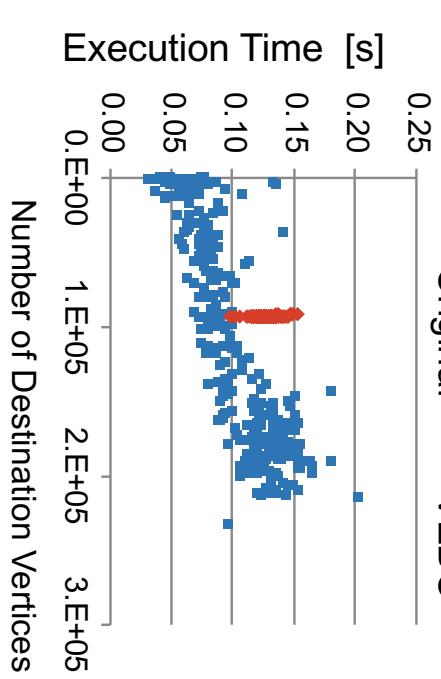
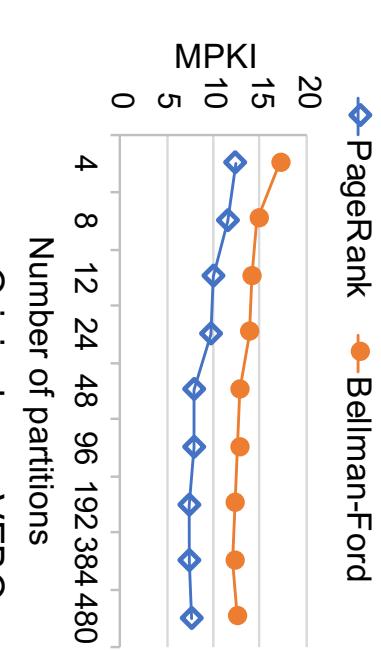
- Solution:

- Partition graphs finely: improves memory locality, unit of work for scheduling, load balancing
- Co-locate partitions and the main code operating on them on the same NUMA node
- Vertex and Edge Balanced Reordering (VETO):
 - Relabel vertices in order to have $\#vertices/partition$ AND same $\#edges/partition$



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Memory accesses per 1000 instructions (MPKI)
twitter graph



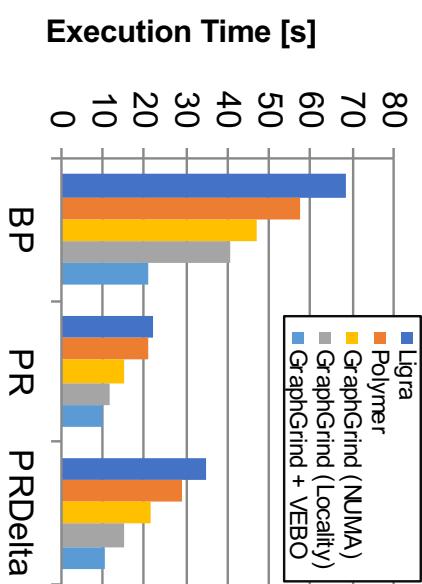
GraphGrind: Performance Comparison

Runtime on twitter graph for

BP: Belief Propagation

PR: 10 iterations of PageRank

PRDelta: approximating PageRank



Time per iteration for PageRank on UK-Union 2007
snapshot of internet pages in the .uk domain

Framework	#cores	Time/PageRank iteration
Giraph (Apache)	128	62s
GraphLab (Apache)	128	42s
GraphX (Apache)	128	23s
Ligra (CMU/MIT)	48*	3.8s
GraphGrind	48*	1.2s

*4-socket Xeon E7-4860 v2



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Questions

- The right hardware for the job. What is it?
 - Storage class memory
 - Scale-out or scale-up?
- What are the key challenges in processing industrial graph datasets?
- What workloads are speed-sensitive, which aren't?

