

Distributed Frameworks for Alternating Least Squares

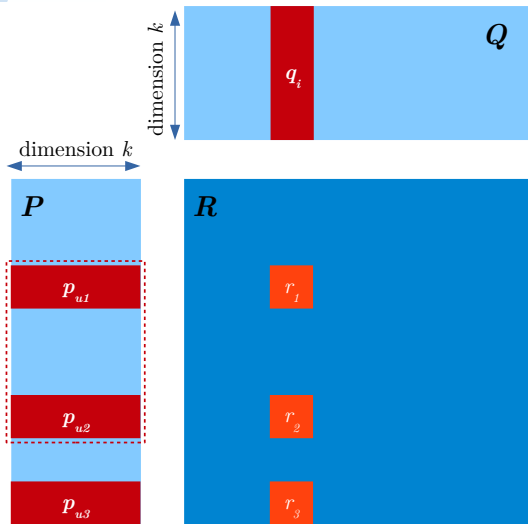
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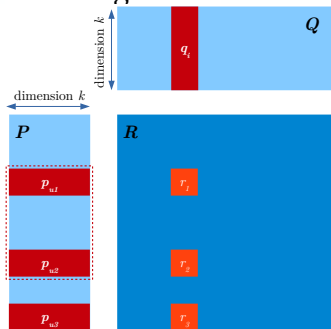
ALTERNATING LEAST SQUARES



ALTERNATING LEAST SQUARES

$$f_{RMSE}(P, Q) = \sum_{(u,i) \in \text{Training}} (R_{ui} - p_u \cdot q_i^T)^2 + \lambda \cdot (\|P\|_F^2 + \|Q\|_F^2)$$

- ▶ Update step for Q : $Q_i \leftarrow (P^T P)^{-1} P^T R_i$
- ▶ For each nonzero rating we communicate $(P^T P)^{-1}$ of dim k^2

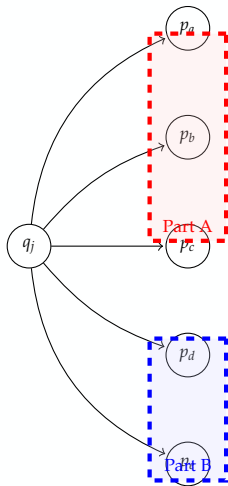


ALS MULTI-MACHINE NO SHARED MEMORY

- ▶ Goal: efficient ALS *and* models for other algorithms
- ▶ Problem: Large amount of communication alternating between rows and columns
 - ALS message size is quadratic in number of latent factors
- ▶ Drawback of "think as a node" philosophy
 - Repeat the same message for all graph nodes
 - Even if they reside on the same server

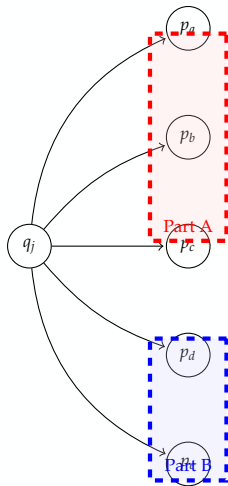
DISTRIBUTION OVERHEAD

- ▶ Partitioned graph or ratings matrix
- ▶ Naive approach: q_j communicates to each p_i individually
- ▶ In ALS, PageRank, ..., messages from q_j are identical
- ▶ **Network communication becomes the bottleneck.**



PROPOSED SOLUTION

- ▶ Efficient communication between partitions
- ▶ Translated to graph processing this is just a *multicast*.



BIG DATA FRAMEWORKS

- ▶ Big Data frameworks lack an operator for this job.
 - Hadoop (Mahout) Map, Reduce
 - Spark “Functional” operators on (memory) Resilient Distributed Datasets
 - Flink “Functional” operators and iteration
 - Our experimental platform
- ▶ **Notion of the partition hidden from user when implementing ALS by vector-to-vector communication.**

BIG DATA FRAMEWORKS - SOLUTION

- ▶ Mahout implementation: “CustomALS”.
- ▶ Algorithm provides an artificial partition ID
- ▶ Map-Reduce grouped by partition ID, expected one partition per reducer
- ▶ Partitioning to minimize the communication between partitions **not ensured** but left for the framework

GRAPH PROCESSING ENGINES

- ▶ Bulk Synchronous Parallel (BSP)
 - Sends along ALL nonzero ratings
 - Even if the message is identical
 - This issue holds even for PageRank
- ▶ Example: Giraph
 - “Think like a vertex”, no partition notion
 - No multicast support in framework

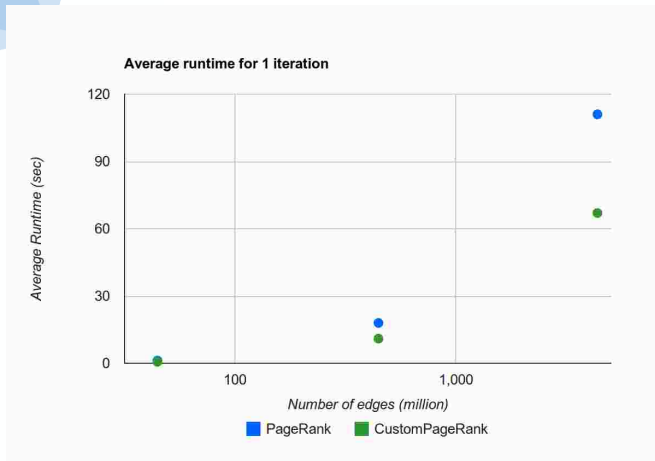
DISTRIBUTED GRAPHLAB

- ▶ Several optimization over plain BSP:
 - Framework support to distribute very high degree nodes: PowerGraph partitions scatters and gathers
 - Optimization: emit unchanged information by caching on gather side
 - Optimization: graph partitioning to reduce number of edges cut (hard to partition a real implicit ratings matrix)
- ▶ **But no handling for multiple identical messages**

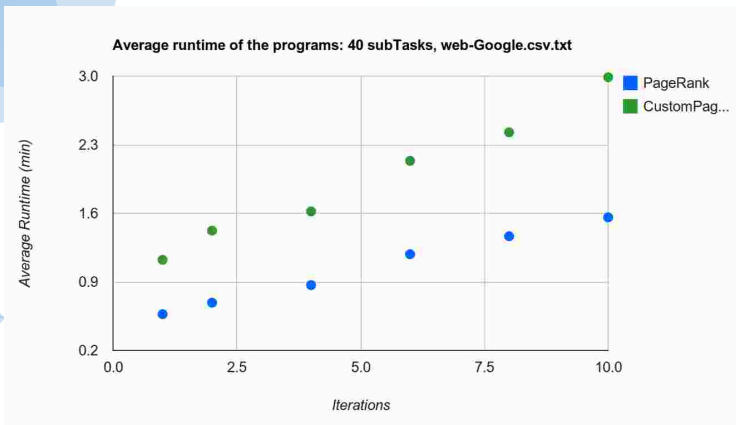
EXPERIMENTS – DISTRIBUTED MESSAGE PASSING IN C++

- ▶ Proof of concept for a low communication task: PageRank
- ▶ We rely on direct control over partitions
- ▶ Each vertex sends the message to relevant partitions once
- ▶ Test on large Web crawl (.pt): 300M nodes, 1B edges
- ▶ Significant improvement

CUSTOM PAGERANK IN C++

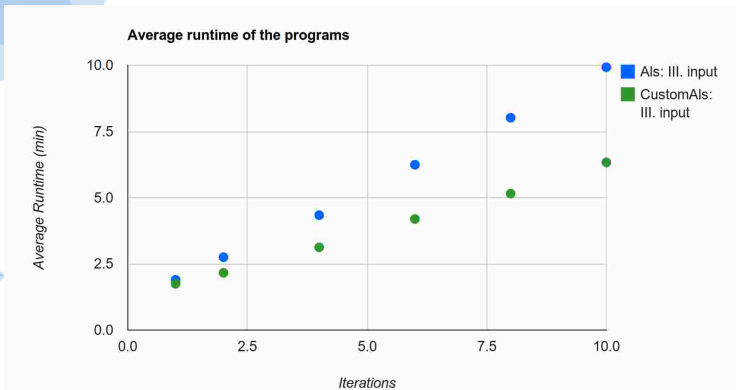


CUSTOMPAGERANK IN APACHE FLINK



- ▶ We define hypernodes – Mahout CustomALS style
- ▶ Insufficient for low communication tasks
- ▶ Web-Google graph from Stanford Large Network Dataset Collection, $9 \cdot 10^5$ nodes,

CUSTOMALS IN APACHE FLINK



- ▶ Generated test data 15 million ratings (courtesy: Gravity)
- ▶ Framework support already sufficient for ALS

CONCLUSIONS

- ▶ ALS multi-machine no shared memory
 - Heavy communication alternating between rows and columns
 - ALS message size is quadratic in number of latent factors
 - Affects MapReduce with no permanent storage (Mahout “CustomALS”)
 - Graph parallel frameworks with nonzero ratings mapped to edges
- ▶ Ongoing experiments with Message Passing, Giraph, Apache Flink, and its Pregel implementation Spargel.
 - Communication primitives to bind identical messages - use multicast
 - Promising even for seemingly low communication intense algorithms such as PageRank.