Why do we need graph processing?





Why do we need graph processing?

- Collaborative Filtering
 - Alternating Least Squares
 - Stochastic Gradient Descent
 - Tensor Factorization
- Structured Prediction
 - Loopy Belief Propagation
 - Max-Product Linear Programs
 - Gibbs Sampling
- Semi-supervised ML
 - Graph SSL

- CoEM
- Community Detection
 - Triangle-Counting
 - K-core Decomposition
 - K-Truss
- Graph Analytics
 - PageRank
 - Personalized PageRank
 - Shortest Path
 - Graph Coloring

Many of these can be expressed as matrix problems





The Apache Software Foundation Dashboards - Projects Issues 🔻 Agile ttp://www.apache.org Spark / SPARK-3789 Spark X] Python bindings for GraphX Agile Board Details Type: New Feature Status: Priority: Maior Resolution: Unresolved Affects Version/s: Fix Version/s: None None GraphX, PySpark Component/s: Labels: None Attachments PyGraphX_design_doc.pdf 161 kB Issue Links SPARK-3665 Java API for GraphX

"The entire world of ecommerce on the internet is driven by graph **analytics** (page rank, suggestions, also viewed, etc. etc.) While arcane to some, is a very important and growing field of computer science and web analytics. ESPECIALLY where big data is concerned."

requires

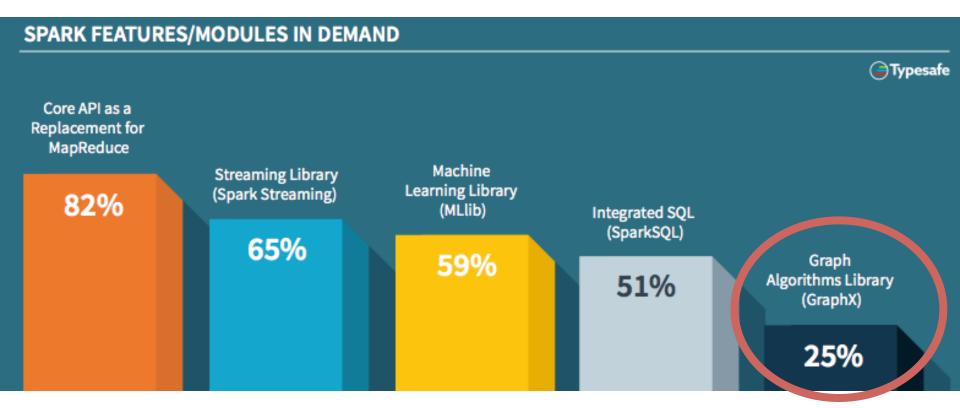
"Given the lack of activity on GraphX and its defunct predecessor Bagel, I doubt anything significant will be added here. I'd almost just close this."

"My personal bias is that most real-world problems that look like they'd be cool to solve as graph problems, aren't graph problems or aren't great to actually solve that way"

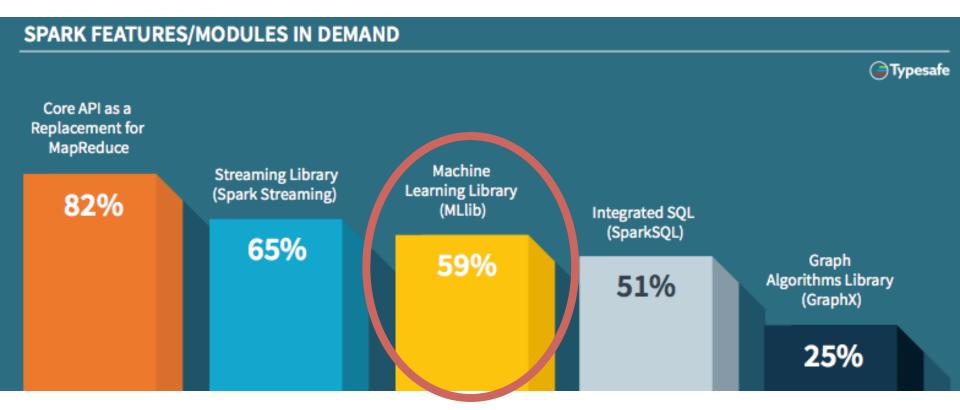
"FWIW virtually none of our customers use GraphX, and we interact with a pretty good cross section of Big Companies. Many of the useful functions you identify are not solved as graph problems in my experience, even if they could be (e.g. recommenders, also viewed)."

Can we infer from the comments on this ticket that GraphX will be discontinued and no longer supported by the Spark Community? I see the makings of rumors already...

Which of the following Spark features or modules are most likely to solve your big data challenges? (Typesafe + Databricks + Dzone surveyed 2136 people)



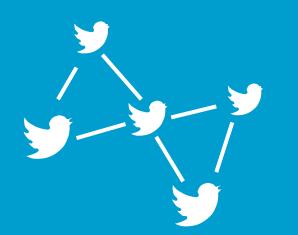
Some Mllib algorithms use GraphX: Power Iteration Clustering, LDA

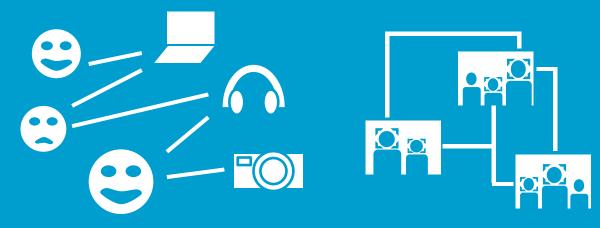


Trends

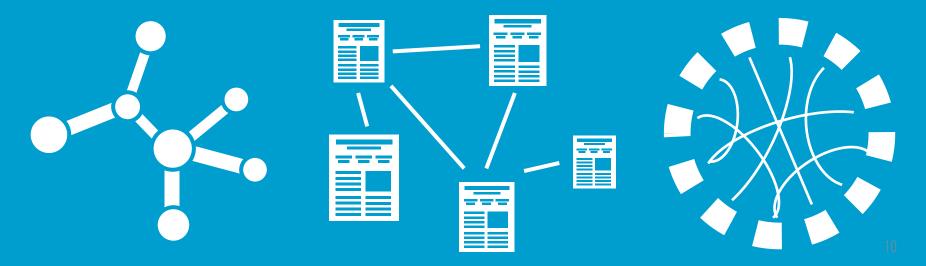
Why do we need **distributed** graph processing?

- Graphs used in GraphX paper have billions of edges
 Twitter: 40m users, 1.4billion links
- Frank McSherry's laptop can process them faster than GraphX





Billions of Edges Rich Metadata

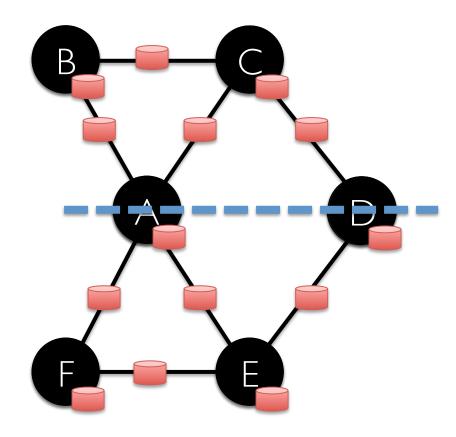


Why do we need **distributed** graph processing?

- To process graphs with lots of metadata
 Is the metadata needed for the graph problem?
- Because it's convenient to incorporate in a single system
 Include fast single-machine implementation (as fallback) in Spark?

What's hard about distributed graph processing?

- How do you represent the graph?
- How do you distribute the graph over the machine?
 - Some parts of the graph need to be duplicated
 - Existing systems: specialized representation

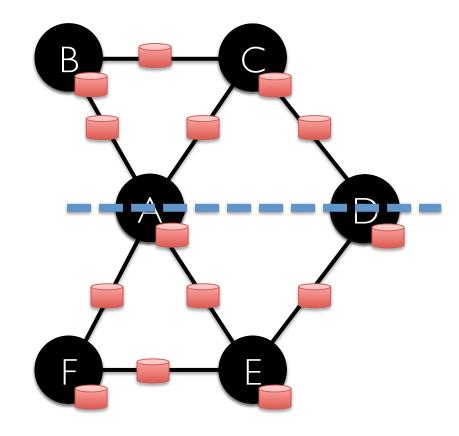


GraphX: can we use a general-purpose system?

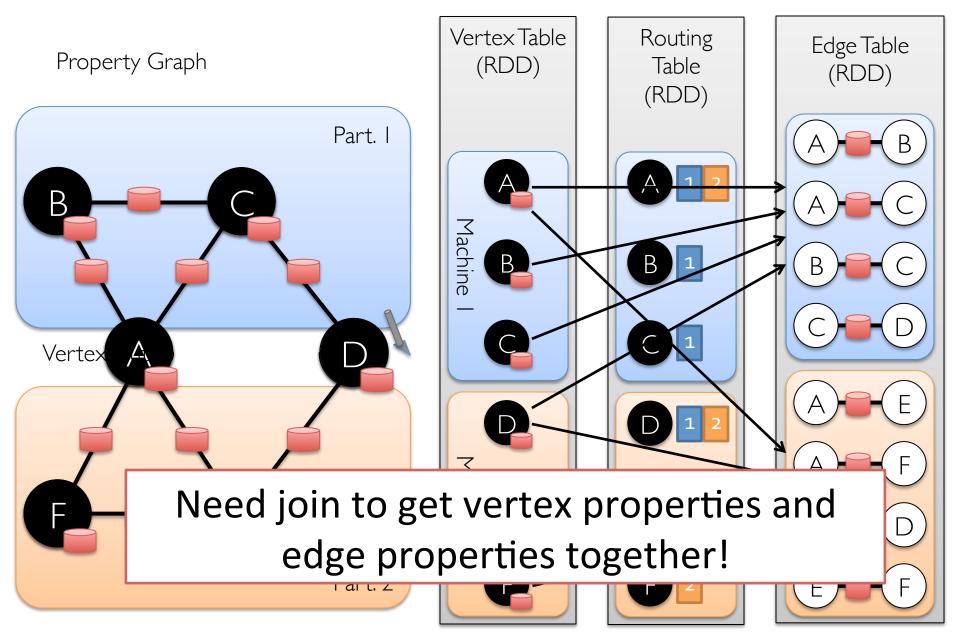
- Graph computation often part of a bigger pipeline
- Why now?
 - Frameworks support in-memory processing (kind of)
 - Allow fine-grained control over data partitioning

Fundamental challenge: data representation

Dataflow systems expect a single, partitioned dataset



Encoding Property Graphs as Tables



What changed?

Graph Processing Systems

- Pregel: synchronous steps
- GraphLab: asynchronous steps
- PowerGraph: better placement / representation
- GraphX: built on general purpose system, synchronous

Takeaway: Spark as a building block

- Gave control over data storage (memory / disk)
- Gave control over data partitioning
- Doesn't support asynchrony



Existing paper says: network doesn't matter for GraphX

I optimized the CPU time of PageRank

Now the network matters more

Why are Frank McSherry's things so much faster than GraphX?

- His laptop was faster
- Timely dataflow was much faster
- Cost of generality?
 - With simple types, GraphX spends much of its time boxing/ unboxing primitive types
 - Serialization is not optimized
 - Spark is in the process of improving this