Cloud Computing CS 15-319

Dryad and GraphLab Lecture 11, Feb 22, 2012

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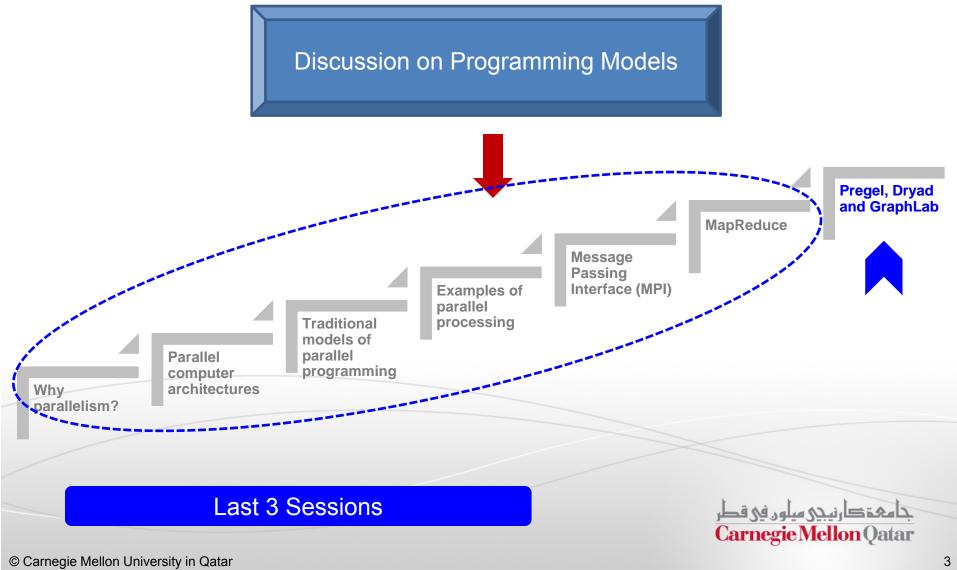
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- Last session
 - Pregel
- Today's session
 - Dryad and GraphLab
- Announcement:
 - Project Phases I-A and I-B are due today



Objectives







- In this part, the following concepts of Dryad will be described:
 - Dryad Model
 - Dryad Organization
 - Dryad Description Language and An Example Program
 - Fault Tolerance in Dryad



In this part, the following concepts of Dryad will be described:

Dryad Model

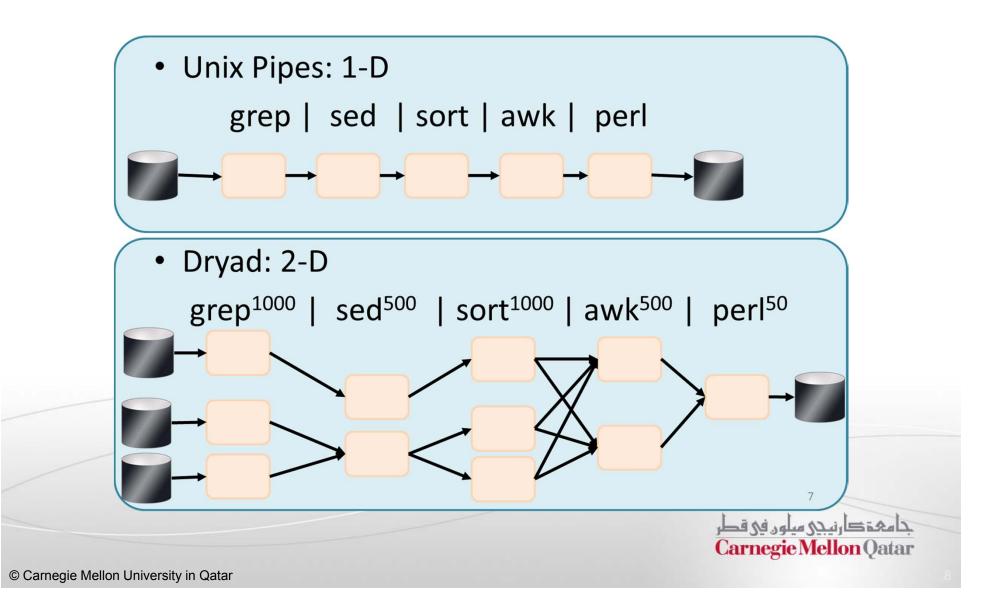
- Dryad Organization
- Dryad Description Language and An Example Program
- Fault Tolerance in Dryad

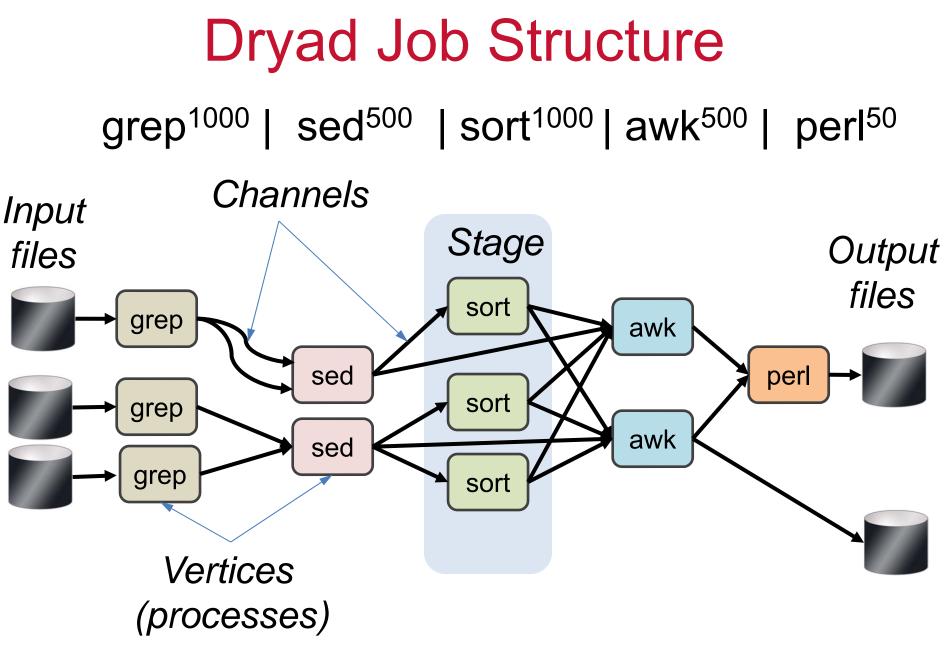


- Dryad is a general purpose, high-performance, distributed computation engine
- Dryad is designed for:
 - High-throughput
 - Data-parallel computation
 - Use in a private datacenter
- Computation is expressed as a directed-acyclic-graph (DAG)
 - Vertices represent programs
 - Edges represent data channels between vertices

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Unix Pipes vs. Dryad DAG



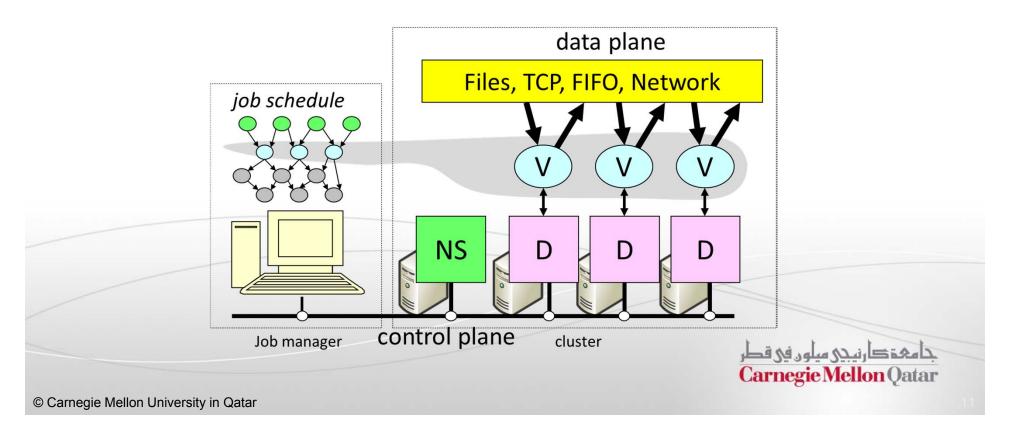


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 - Fault Tolerance in Dryad



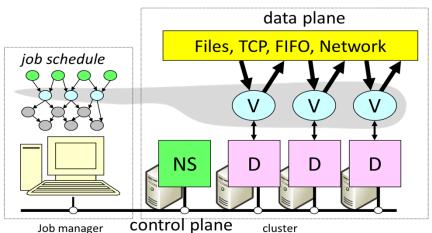
Dryad System Organization

- There are 3 roles for machines in Dryad
 - Job Manager (JM)
 - Name Server (NS)
 - Daemon (D)



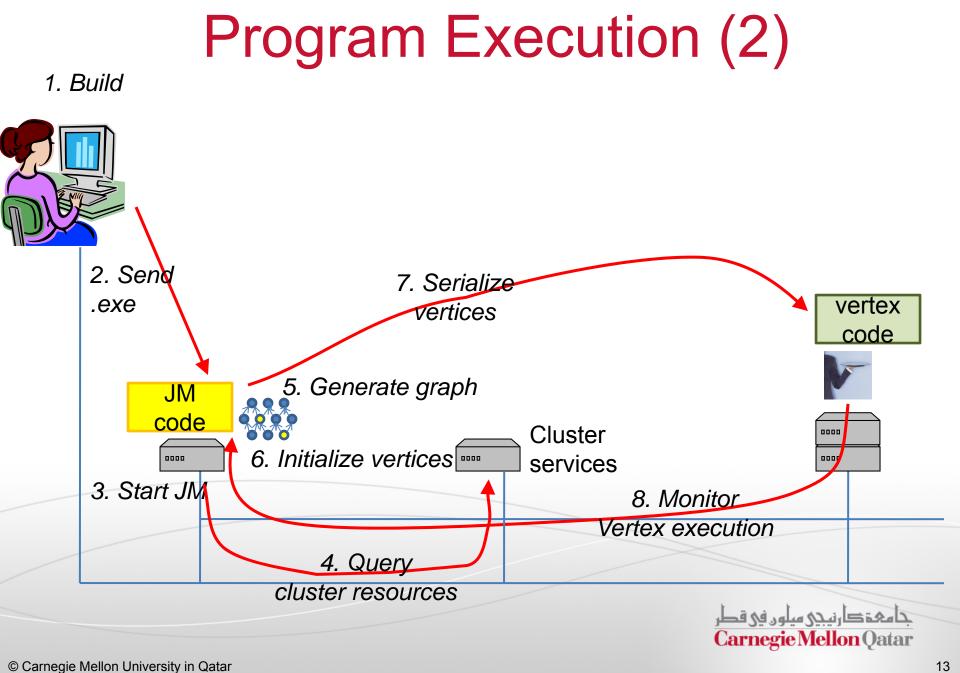
Program Execution (1)

- The Job Manager (JM):
 - Creates the job communication graph (job schedule)
 - Contacts the *NS* to determine the number of *D*s and the topology
 - Assigns Vs to each D (using a simple task schedulernot described) for execution
 - Coordinates data flow through the data plane



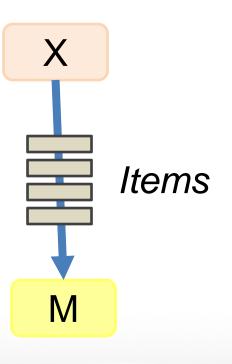
- Data is distributed using a distributed storage system that shares with the Google File System some properties (e.g., data are split into chunks and replicated across machines)
- Dryad also supports the use of NTFS for accessing files locally

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Data Channels in Dryad

- Data items can be *shuffled* between vertices through data channels
- Data channels can be:
 - Shared Memory FIFOs (intra-machine)
 - TCP Streams (inter-machine)
 - SMB/NTFS Local Files (temporary)
 - Distributed File System (persistent)
- The performance and fault tolerance of these mechanisms vary
- Data channels are abstracted for maximum flexibility



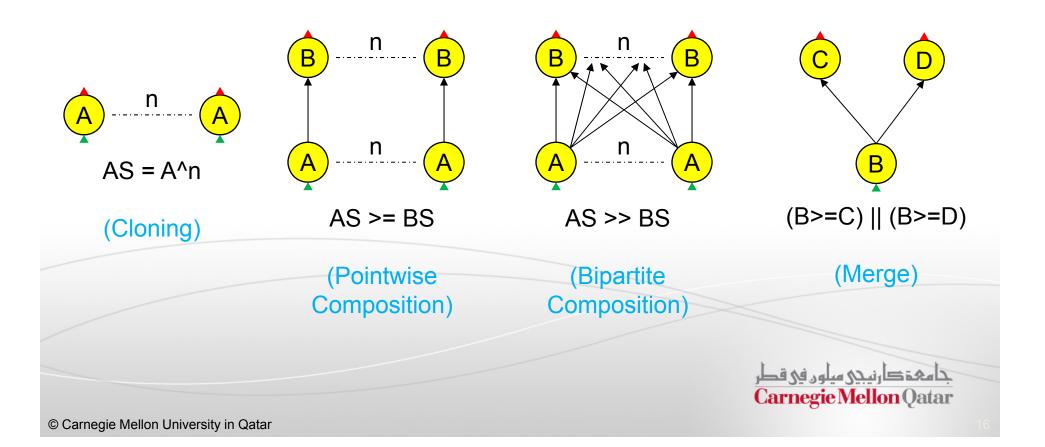
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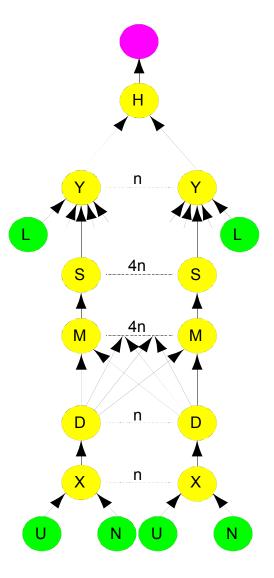
Dryad Graph Description Language

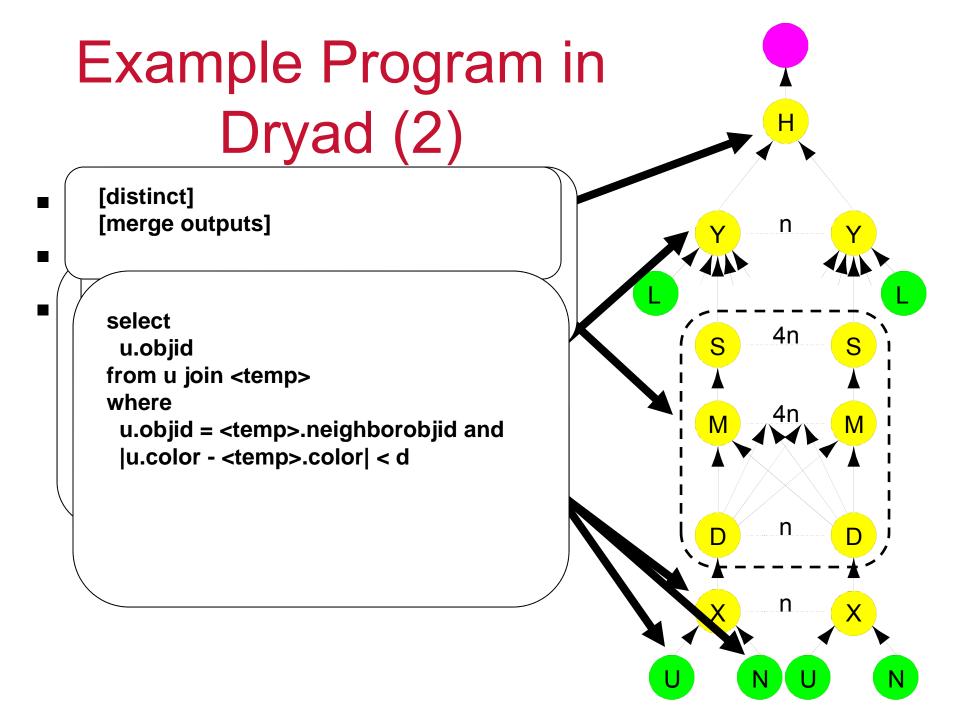
Here are some operators in the Dryad graph description language:



Example Program in Dryad (1)

- Skyserver SQL Query (Q18):
 - Find all the objects in the database that have neighboring objects within 30 arc seconds such that at least one of the neighbors has a color similar to the primary object's color
 - There are two tables involved
 - photoObjAll and it has 354,254,163 records
 - Neighbors and it has 2,803,165,372 records
 - For the equivalent Dryad computation, they extracted the columns of interest into two binary files, "ugriz.bin" and "neighbors.bin"

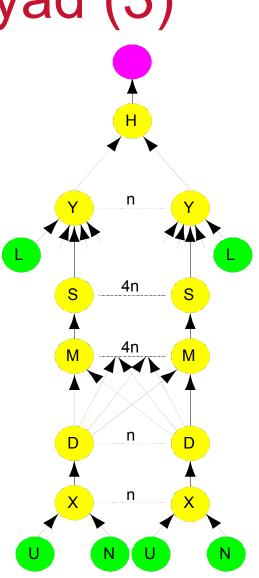




Example Program in Dryad (3)

Here is the corresponding Dryad code:

```
GraphBuilder XSet = moduleX<sup>N</sup>;
GraphBuilder DSet = moduleD<sup>N</sup>;
GraphBuilder MSet = moduleM<sup>^</sup>(N*4);
GraphBuilder SSet = moduleS<sup>(N*4)</sup>;
GraphBuilder YSet = moduleY<sup>N</sup>;
GraphBuilder HSet = moduleH<sup>1</sup>;
GraphBuilder XInputs = (ugriz1 >= XSet) || (neighbor >= XSet);
GraphBuilder YInputs = ugriz2 >= YSet;
GraphBuilder XToY = XSet >= DSet >> MSet >= SSet;
for (i = 0; i < N*4; ++i)
Ł
XToY = XToY || (SSet.GetVertex(i) >= YSet.GetVertex(i/4));
}
GraphBuilder YToH = YSet >= HSet;
GraphBuilder HOutputs = HSet >= output;
GraphBuilder final = XInputs || YInputs || XToY || YToH ||
HOutputs;
```

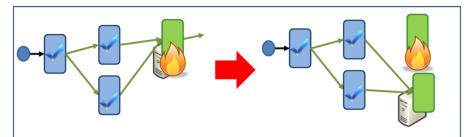


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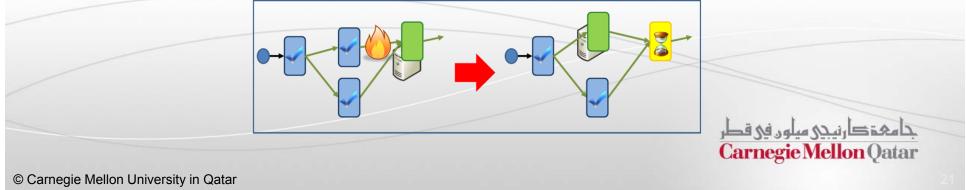


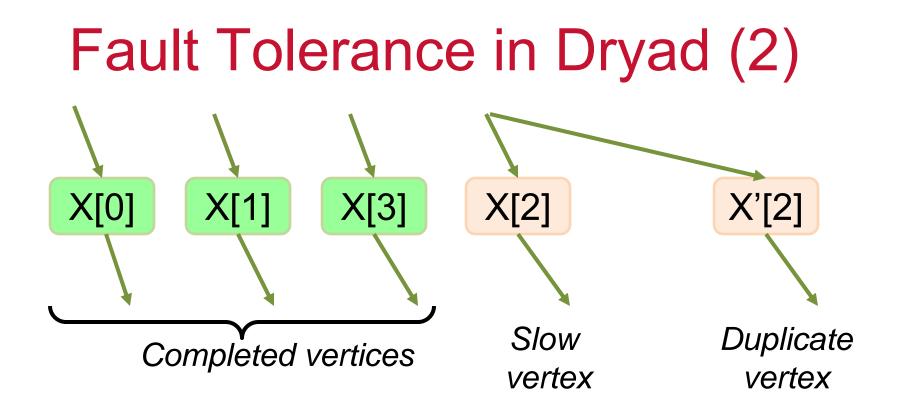
Fault Tolerance in Dryad (1)

- Dryad is designed to handle two types of failures:
 - Vertex failures
 - Channel failures
- Vertex failures are handled by the JM and the failed vertex is re-executed on another machine



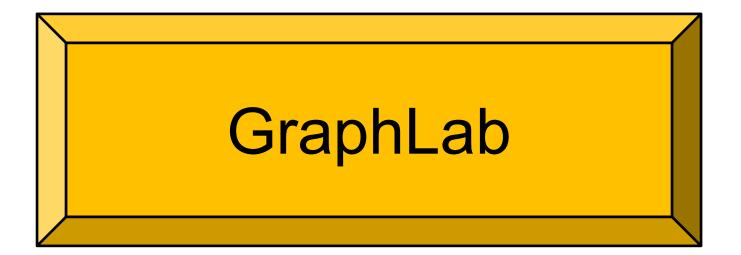
Channel failures cause the preceding vertex to be re-executed





Duplication Policy = f(running times, data volumes)

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GraphLab

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 - Motivation for GraphLab
 - GraphLab Data Model and Update Mechanisms
 - Scheduling in GraphLab
 - Consistency Models in GraphLab
 - PageRank in GraphLab



GraphLab

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Motivation for GraphLab

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Motivation for GraphLab

- Shortcomings of MapReduce
 - Interdependent data computation difficult to perform
 - Overheads of running jobs iteratively disk access and startup overhead
 - Communication pattern is not user definable/flexible
- Shortcomings of Pregel
 - BSP model requires synchronous computation
 - One slow machine can slow down the entire computation considerably

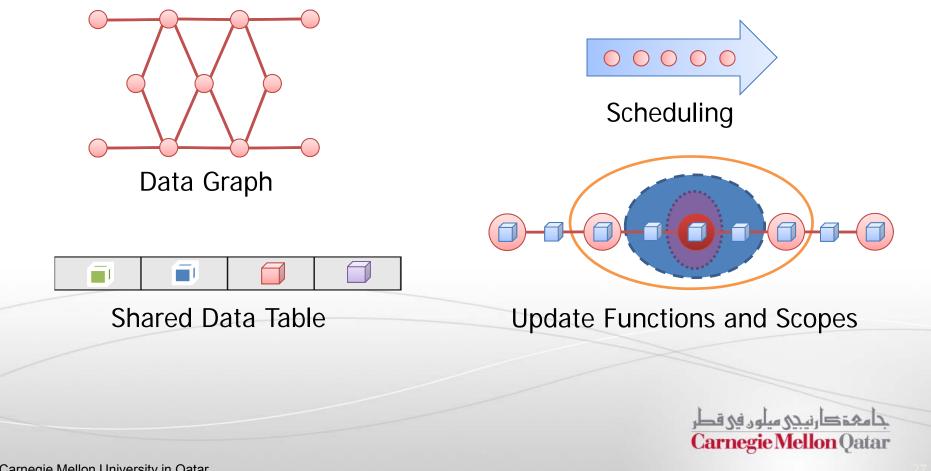
Shortcomings of Dryad

Very flexible but steep learning curve for the programming model

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GraphLab

GraphLab is a framework for parallel machine learning



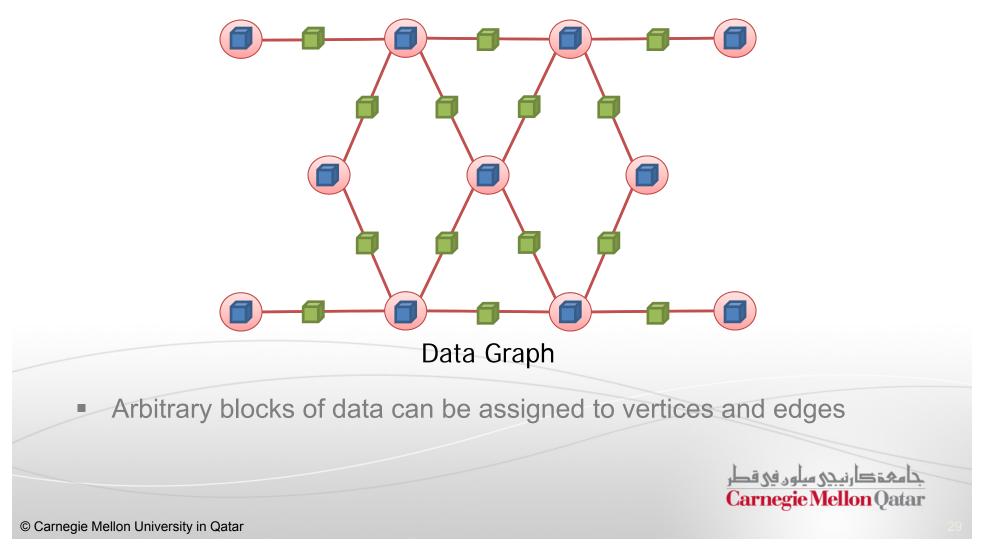
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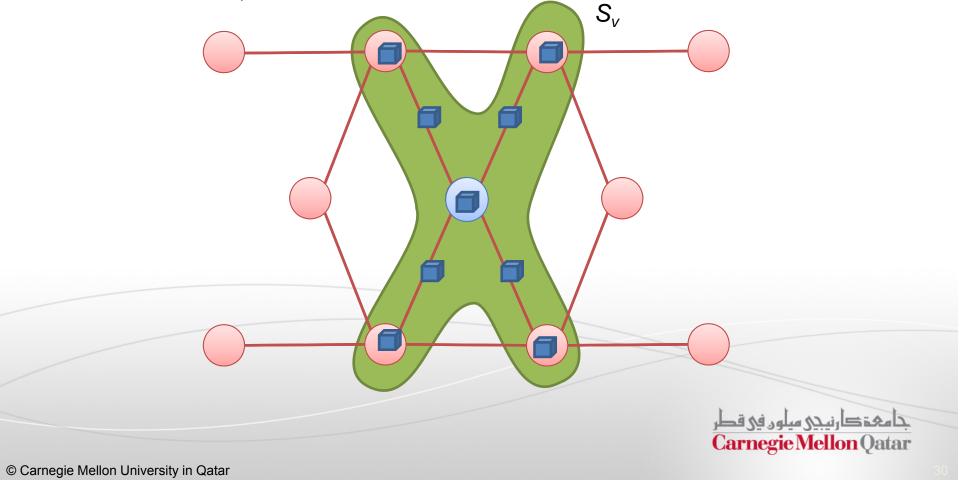
Data Graph

• A graph in GraphLab is associated with data at every vertex and edge



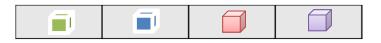
Update Functions

- The data graph is modified using update functions
 - The update function can modify a vertex v and its neighborhood, defined as the scope of v (S_v)



Shared Data Table

- Certain algorithms require global information that is shared among all vertices (Algorithm Parameters, Statistics, etc.)
 - GraphLab exposes a Shared Data Table (SDT)
- SDT is an associative map between keys and arbitrary blocks of data
 - $T[Key] \rightarrow Value$



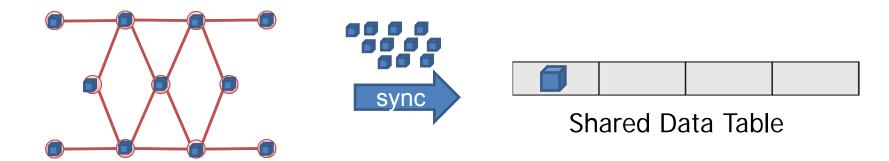
Shared Data Table

The shared data table is updated using the sync mechanism

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Sync Mechanism

- Similar to Reduce in MapReduce
 - User can define fold, merge and apply functions that are triggered during the global sync mechanism
- Fold function allows the user to sequentially aggregate information across all vertices
- Merge optionally allows user to perform a parallel tree reduction on the aggregated data collected during the fold operation
- Apply function allows the user to finalize the resulting value from the fold/merge operations (such as normalization etc.)



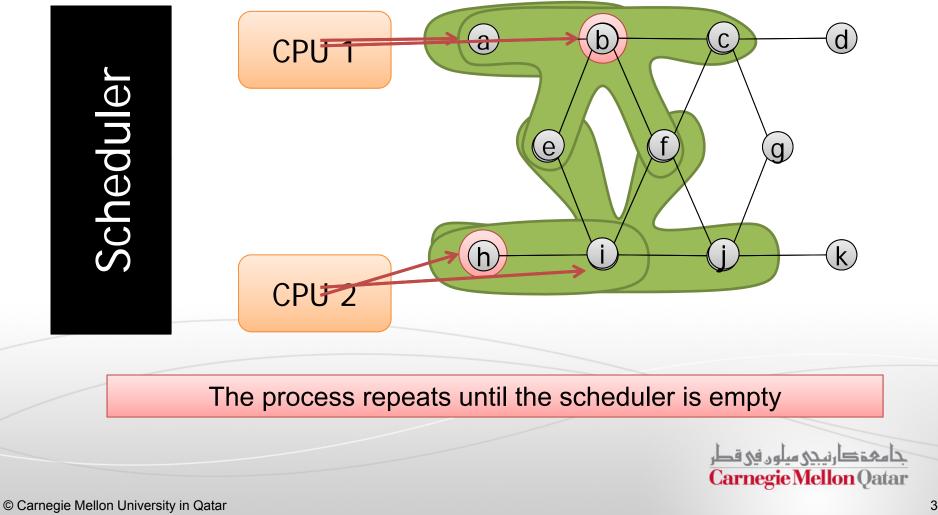
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Scheduling in GraphLab (1)

The scheduler determines the order that vertices are updated



Scheduling in GraphLab (2)

- An update schedule defines the order in which update functions are applied to vertices
 - A parallel data-structure called the scheduler represents an abstract list of tasks to be executed in Graphlab
- Base (Vertex) schedulers in GraphLab
 - Synchronous scheduler
 - Round-robin scheduler
- Job Schedulers in GraphLab
 - FIFO scheduler
 - Priority scheduler
- Custom schedulers can be defined by the set scheduler
- Termination Assessment
 - If the scheduler has no remaining tasks
 - Or, a termination function can be defined to check for convergence in the data



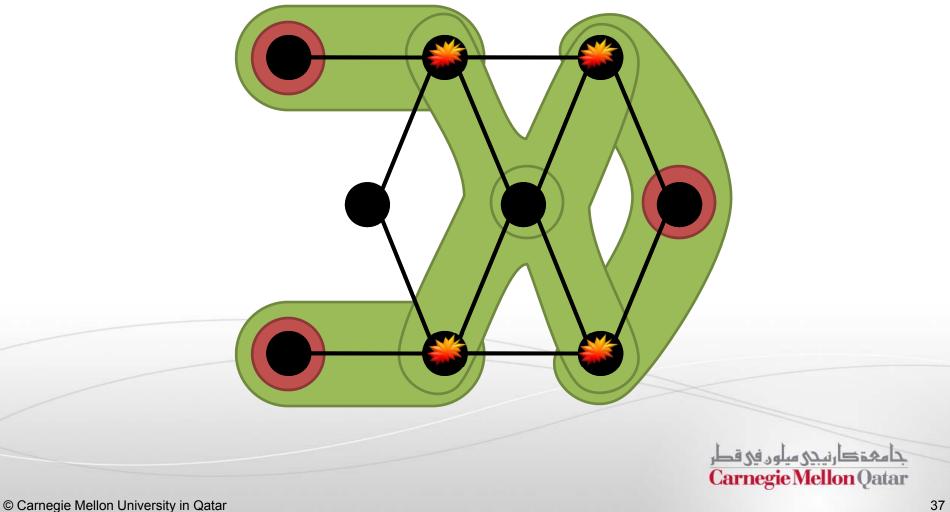
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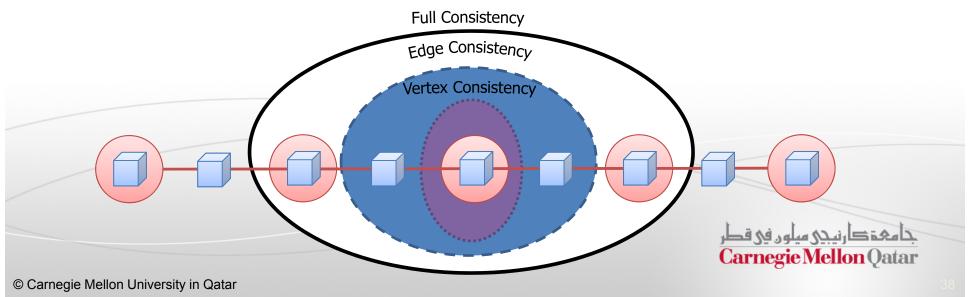
Need for Consistency Models

How much can computation overlap?



Consistency Models in GraphLab

- GraphLab guarantees sequential consistency
 - Guaranteed to give the same result as a sequential execution of the computational steps
- User-defined consistency models
 - Full Consistency
 - Vertex Consistency
 - Edge Consistency



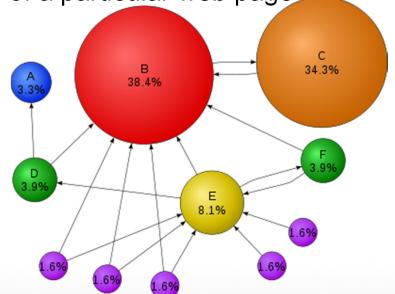
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PageRank (1)

- PageRank is a link analysis algorithm
- The rank value indicates an importance of a particular web page
- A hyperlink to a page counts as a vote of support
- A page that is linked to by many pages with high PageRank receives a high rank itself



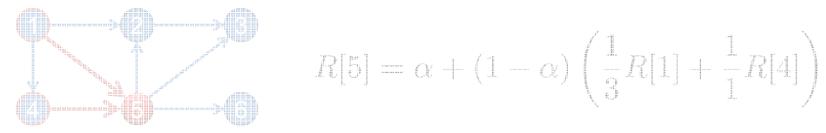
 A PageRank of 0.5 means there is a 50% chance that a person clicking on a random link will be directed to the document with the 0.5 PageRank

PageRank (2)

Iterate:

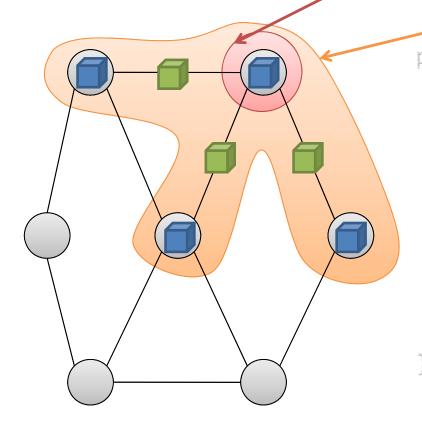
$$R[i] = \alpha + (1 - \alpha) \sum_{(j,i) \in E} \frac{1}{L[j]} R[j]$$

- Where:
 - α is the random reset probability
 - L[j] is the number of links on page j



PageRank Example in GraphLab

 PageRank algorithm is defined as a per-vertex operation working on the scope of the vertex



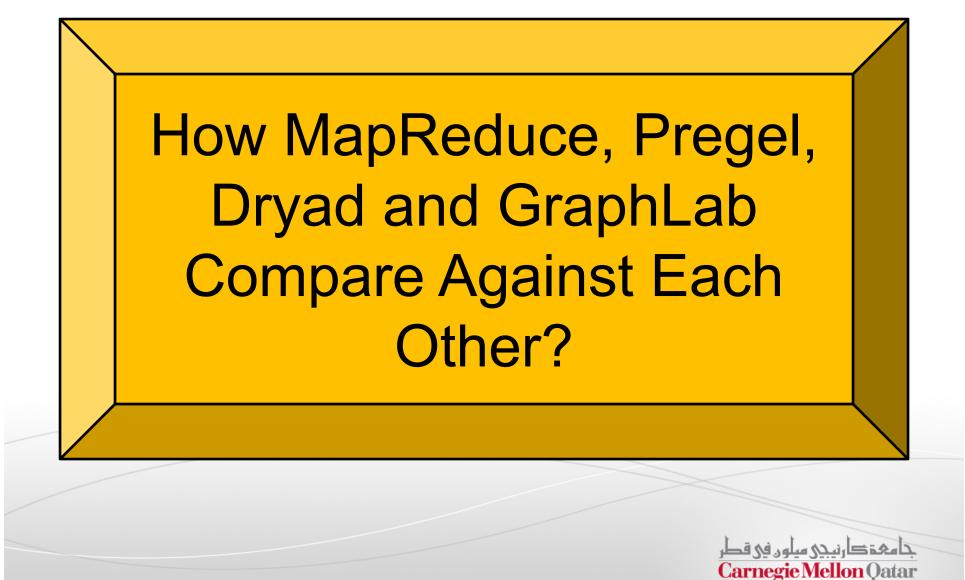
pagerank(i, scope){
 // Get Neighborhood data
 (R[i], W_{ij}, R[j]) <scope;</pre>

// Update the vertex data

$$R[i] \leftarrow \alpha + (1 - \alpha) \sum_{j \in N[i]} W_j \times R[j];$$

// Reschedule Neighbors if needed
if R[i] changes then
 reschedule_neighbors_of(i);

Dynamic computation



Comparison of the Programming Models

	MapReduce	Pregel	Dryad	GraphLab
Programming Model	Fixed Functions – Map and Reduce	Supersteps over a data graph with messages passed	DAG with program vertices and data edges	Data graph with shared data table and update functions
Parallelism	Concurrent execution of tasks within map and reduce phases	Concurrent execution of user functions over vertices within a superstep	Concurrent execution of vertices during a stage	Concurrent execution of non-overlapping scopes, defined by consistency model
Data Handling	Distributed file system	Distributed file system	Flexible data channels: Memory, Files, DFS etc.	Undefined – Graphs can be in memory or on disk
Task Scheduling	Fixed Phases – HDFS Locality based map task assignment	Partitioned Graph and Inputs assigned by assignment functions	Job and Stage Managers assign vertices to available daemons	Pluggable schedulers to schedule update functions
Fault Tolerance	DFS replication + Task reassignment / Speculative execution of Tasks	Checkpointing and superstep re- execution	Vertex and Edge failure recovery	Synchronous and asychronous snapshots
Developed by	Google	Google	Microsoft	Carnegie Mellon

References

- This presentation has elements borrowed from various papers and presentations:
- Papers:
 - Pregel: <u>http://kowshik.github.com/JPregel/pregel_paper.pdf</u>
 - Dryad: <u>http://research.microsoft.com/pubs/63785/eurosys07.pdf</u>
 - GraphLab: <u>http://www.select.cs.cmu.edu/publications/paperdir/uai2010-low-gonzalez-kyrola-bickson-guestrin-hellerstein.pdf</u>
- Presentations:
 - Dryad Presentation at Berkeley by M. Budiu: <u>http://budiu.info/work/dryad-talk-berkeley09.pptx</u>
 - GraphLab1 Presentation: <u>http://graphlab.org/uai2010_graphlab.pptx</u>
 - GraphLab2 Presentation: <u>http://graphlab.org/presentations/nips-biglearn-2011.pptx</u>

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Next Class

Distributed File Systems

