

Flexible Cluster Computing: Dryad and DryadLINQ

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presented by Michael George

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Computing on Clusters

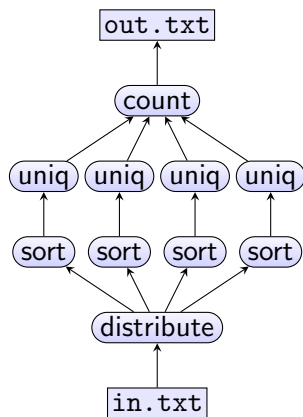
How to crunch lots of data?

- ▶ Explicit distribution
 - ▶ Write it by hand
 - ▶ Hard! failure, resource allocation, scheduling, ...
- ▶ Implicit distribution
 - ▶ MapReduce, DryadLINQ
 - ▶ Easy! As long as your computation is expressible...
- ▶ Virtualized distribution
 - ▶ Dryad
 - ▶ In between. Programmer specifies data flow, system handles details

Dryad Overview — Jobs

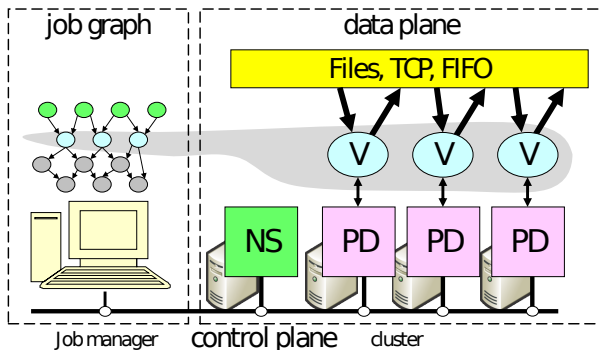
Dryad Job is a Directed Acyclic Graph

- ▶ Vertices are subcomputations
- ▶ Edges are data channels
- ▶ Graph is *virtual* — may be more or fewer vertices than cluster nodes.



Dryad Overview — Execution

Centralized *Job Manager* distributes virtual graph to actual cluster



Writing Vertex Programs

- ▶ A Vertex Program is a class that extends the base VP class
 - ▶ Base class provides typed I/O channels
 - ▶ Specialized abstract subclasses available
 - ▶ Map, Reduce, Distribute, ...
- ▶ Special support for legacy executables
 - ▶ `grep`, `perl`, `legacyApp`, ...
- ▶ Asynchronous I/O API available for vertices that require it
 - ▶ Runtime distinguishes asynch vertices, executes them efficiently on thread pool

Composing Vertex Programs

Vertex programs are joined into graphs

- ▶ edges are local files by default
- ▶ can also be TCP pipes or in-memory FIFOs

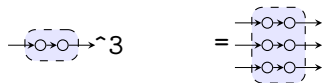
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Predefined operators for common composition patterns:

- ▶ Clone ($G \hat{\sim} n$)



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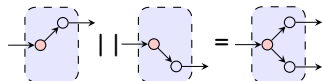
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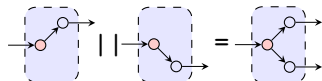
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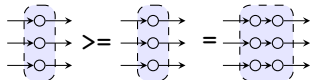
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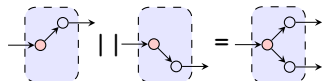
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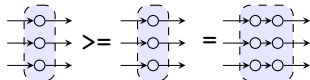
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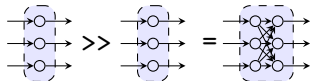
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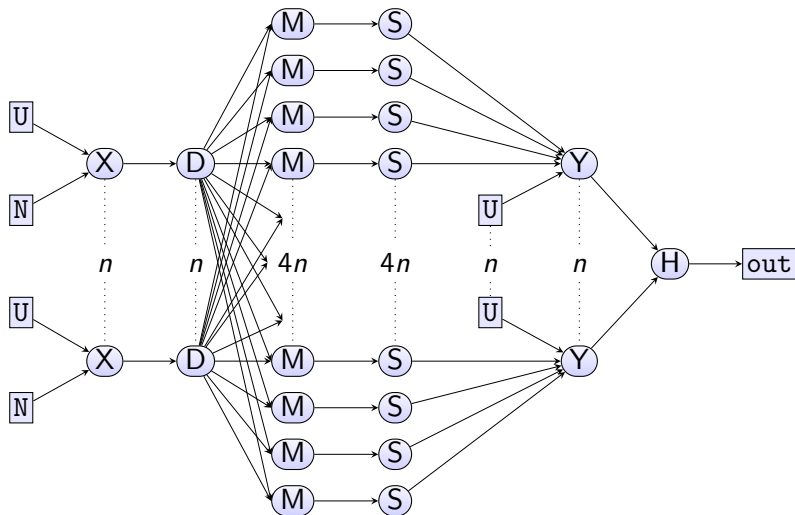
- ▶ Pointwise composition ($G1 \gg G2$)



- ▶ Bipartite composition ($G1 \gg G2$)



Example Job



$((((U \geq X) \parallel (N \geq X)) \geq D)^n \gg ((M \geq S)^4 \gg Y) \parallel (U \geq Y)^n) \gg H \geq out$

Running a Job

Vertices are instantiated on nodes

- ▶ May be multiple *execution records* due to failure
- ▶ Node placement handled by Job Manager
 - ▶ Applications can specify locality “hints” or “requirements”
 - ▶ Edge requirements may force vertices to co-locate

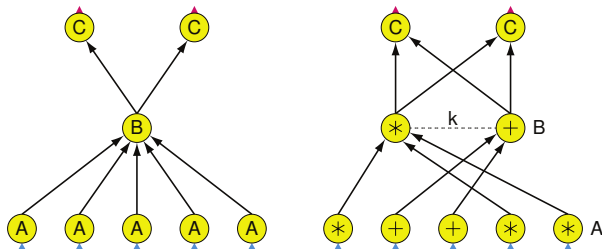
Job manager is notified of node transitions

- ▶ May rerun failed vertices
- ▶ Can rewrite the graph
- ▶ May run duplicate process to route around slow nodes

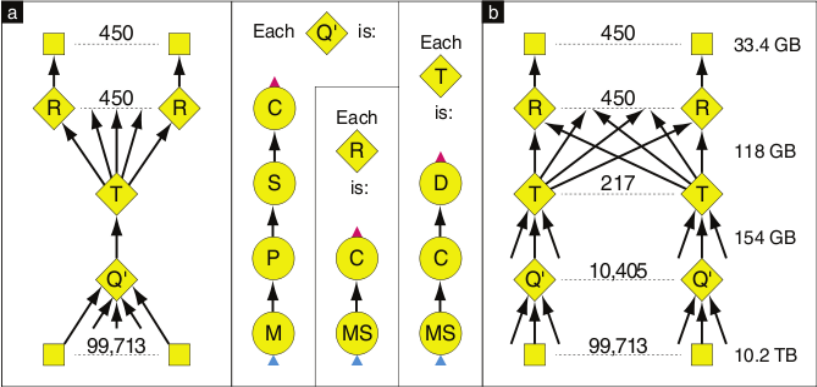
Callback Example: Dynamic Optimization

Client may not know size or distribution of data at load time

- ▶ Can dynamically rewrite the graph



MapReduce in Dryad



So Far...

- ▶ Dryad provides a mid-level execution platform
- ▶ Good performance possible
- ▶ Flexibility
- ▶ But
 - ▶ Data parallelization done by hand
 - ▶ Optimization done by hand
 - ▶ Unfamiliar programming style

DryadLINQ

LINQ (Language INtegrated Query)

- ▶ Standard .NET extension
- ▶ Embeds SQL-like operators into programming languages
- ▶ Developers can mix declarative, functional, and imperative statements

DryadLINQ

- ▶ Compiles LINQ statements to run on Dryad
- ▶ Provides simple, flexible, efficient access to cluster computing

LINQ Example

```
var adjustedScoreTriples =  
    from d in scoreTriples  
    join r in staticRank on d.docID equals r.key  
    select new QueryScoreDecIDTriple(d,r);  
var rankedQueries =  
    from s in adjustedScoreTriples  
    group s by s.query into g  
    select TakeTopQueryResults(g);
```

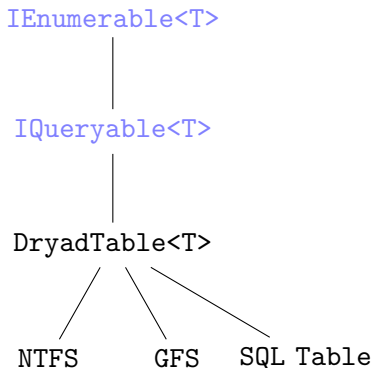
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```
var adjustedScoreTriples =  
    scoreTriples.join(staticRank ,  
        d => d.docID , r => r.key ,  
        (d,r) => new QueryScoreDocIDTriple(d,r));  
var groupedQueries =  
    adjustedScoreTriples.groupBy(s => s.query);  
var rankedQueries =  
    groupedQueries.select(  
        g => TakeTopQueryResults(g));
```

DryadLINQ Constructs

Types:



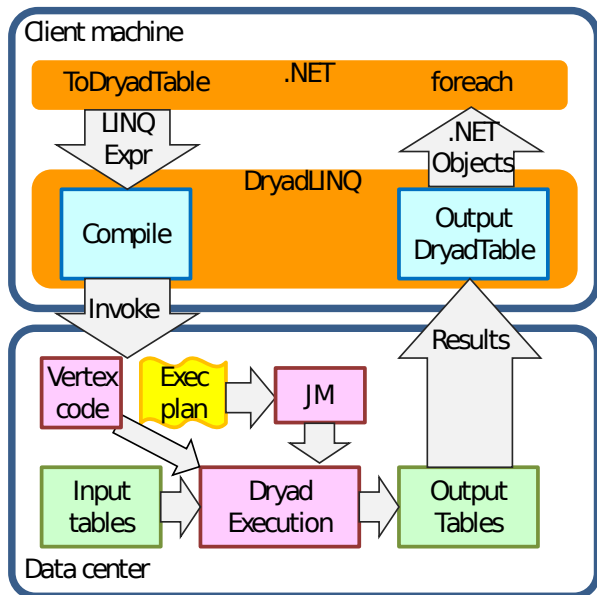
Data Partitioning Operators:

- ▶ `HashPartition<T,K>`
- ▶ `RangePartition<T,K>`

Escape Hatches:

- ▶ `Apply (f)`
- ▶ `Fork (f)`

Execution Overview



Execution Details

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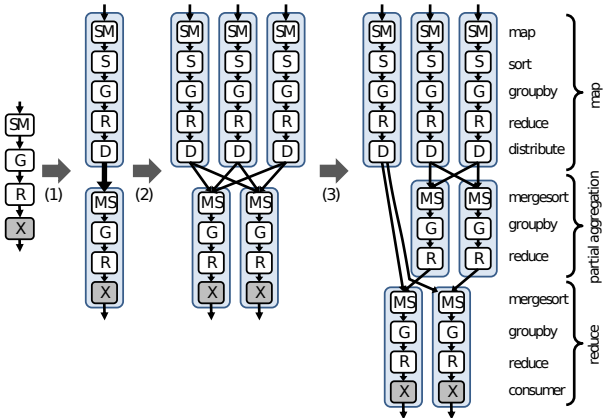
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4. Dynamically, EPG is executed by DryadLINQ job manager
 - ▶ vertices replicated to match data
 - ▶ dynamic optimizations automated
 - ▶ vertices use local LINQ execution engines (e.g. PLINQ)

MapReduce on DryadLINQ

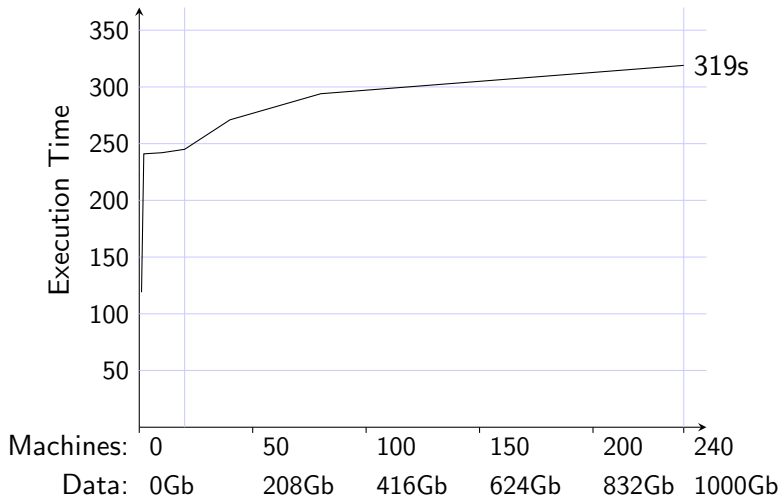
```

public static MapReduce(source, mapper, keySelector, reducer) {
    var mapped = source.selectMany(mapper);
    var groups = mapped.groupBy(keySelector);
    return groups.selectMany(reducer);
}

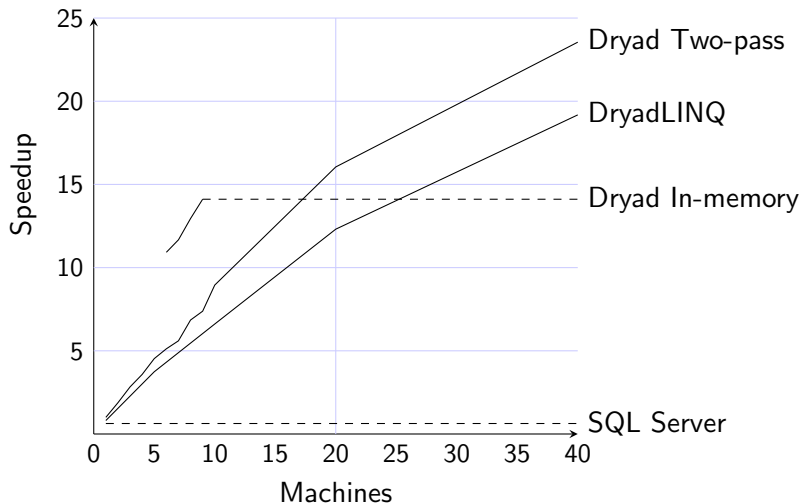
```



Scalability Evaluation — TeraSort



Overhead Evaluation — SkyServer



Conclusions and Discussion

Dryad meets its goals:

- ▶ efficient
- ▶ flexible
- ▶ programmable

DryadLINQ builds on Dryad:

- ▶ almost as efficient
- ▶ concise
- ▶ familiar