# Efficient I/O for Computational Grid Applications

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PhD. Thesis Defense

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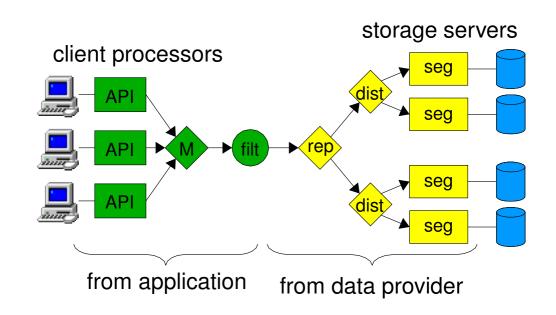
### **Computational Grids**

Networks of geographically distributed heterogeneous systems and devices

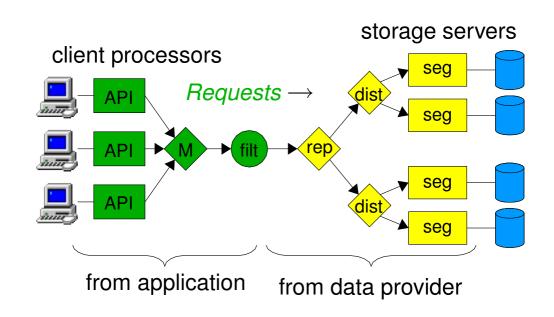
#### Data-intensive scientific applications

- Access large remote datasets (terabyte–petabyte)
- Datasets often need pre/post-processing
- Often computationally intensive
- Examples
  - Climate modeling
  - Astronomy
  - Computational biology
  - High-energy physics

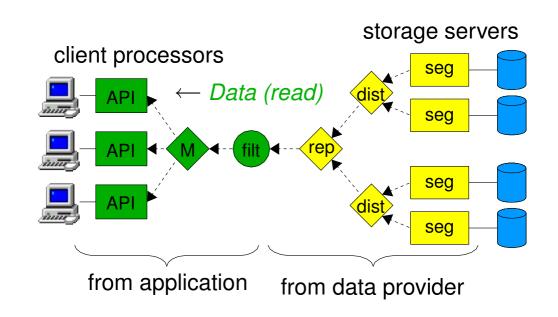
- Application deploys a graph of distributed objects (ships)
- Requests cause pipelined data flow through graph
- Graph has two distinct portions:
  - from the data provider (describes layout of data set)
  - from the application-programmer (pre/post-processing)



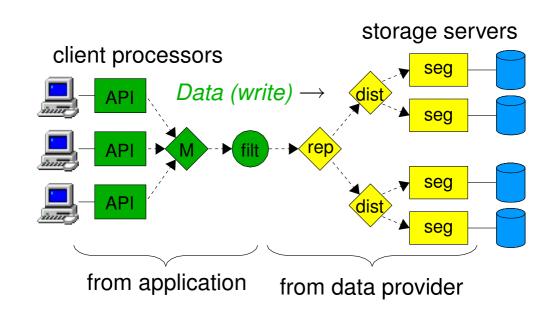
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### Armada

Armada is not a data storage system. Armada is not a parallel file system.

The *data segments* that make up a *data set* are stored in conventional data servers as files, databases, or the like.

The Armada graph encodes most functionality provided by the I/O system:

- programmers interface,
- data layout,
- caching and prefetching policies,
- interfaces to heterogeneous data servers.

### Armada can...

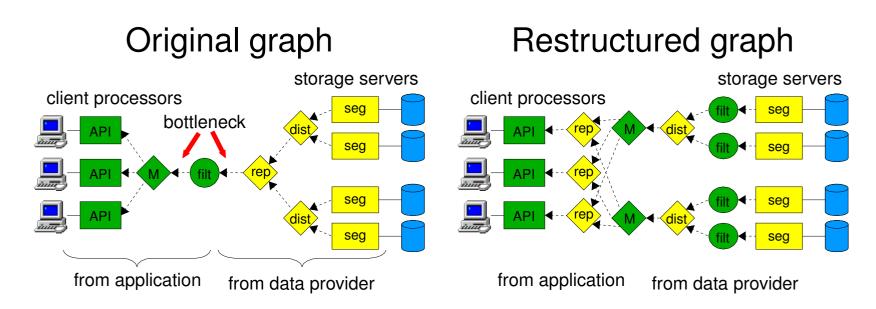
With Armada, one can...

- build a graph for parallel access to a group of legacy files,
- present many similar data sets through a standard interface, and
- provide transparent access to derived "virtual" data either cached or calculated as needed.

### Restructuring

Problems with the example application:

- Potential bottlenecks in composed graph
- original graph restricts placement alternatives for filter

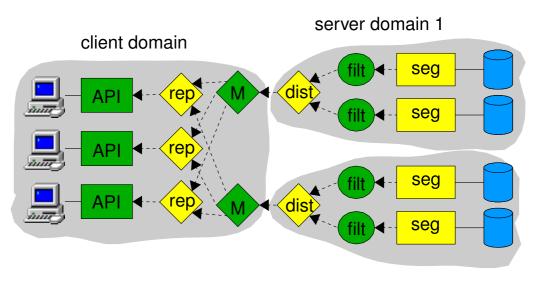


Armada restructures original graph to improve data flow.

### **Placement**

#### After restructuring:

- 1. Armada deploys ships to appropriate administrative domains to optimize data flow, then
- 2. domain-level resource manager decides placement of individual ships.

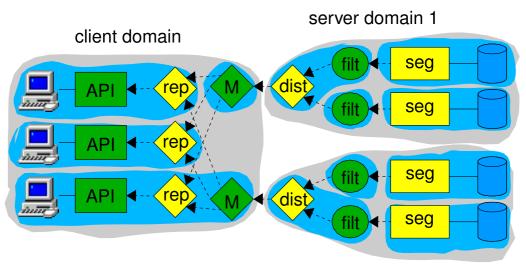


server domain 2

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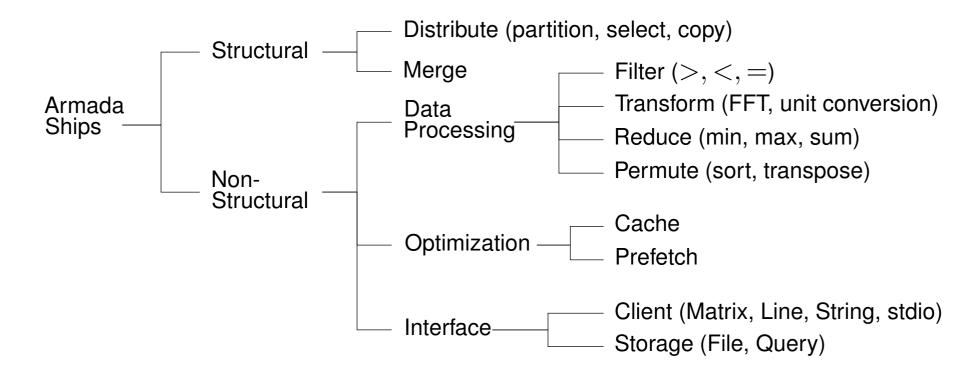


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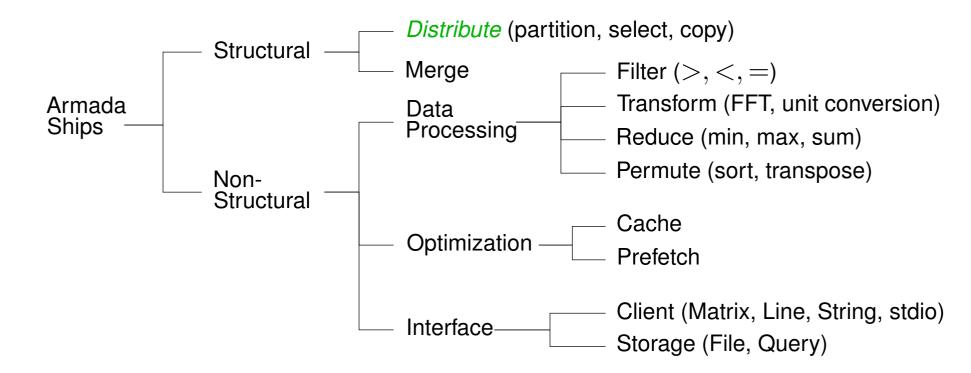
### **Talk Outline**

- Introduction
- Framework details
  - Ships
  - Graph Representation
- Restructuring graphs to improve data flow
- Partitioning graphs and placing ships
- Experiments
- Conclusion

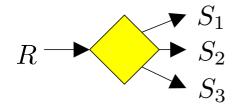
Armada includes a rich set of extensible ship classes.



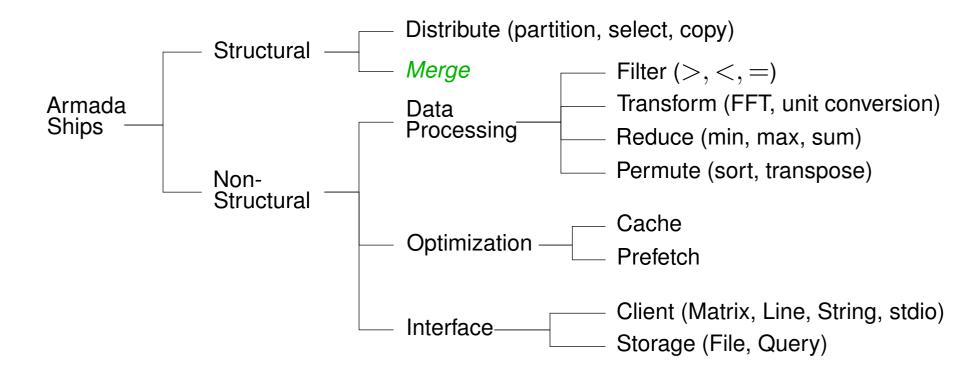
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*Distribute* ships partition requests or data to multiple output streams.



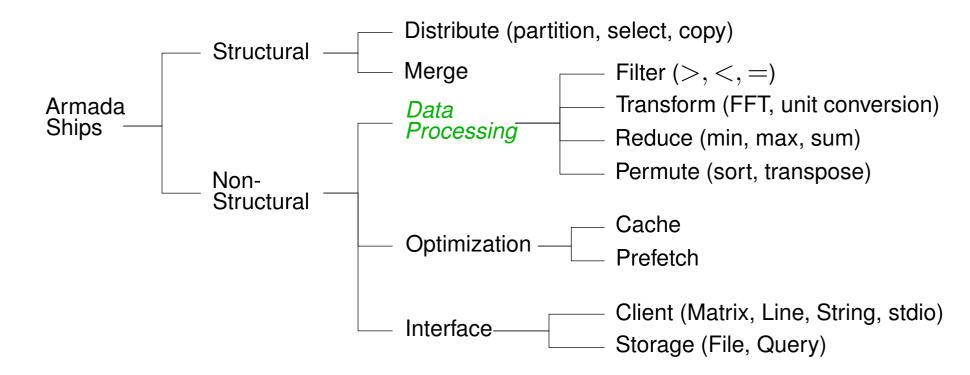
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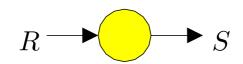
*Merge* ships interleave requests or data from multiple input streams.

$$R_1$$
 $R_2$ 
 $R_3$ 
 $S$ 

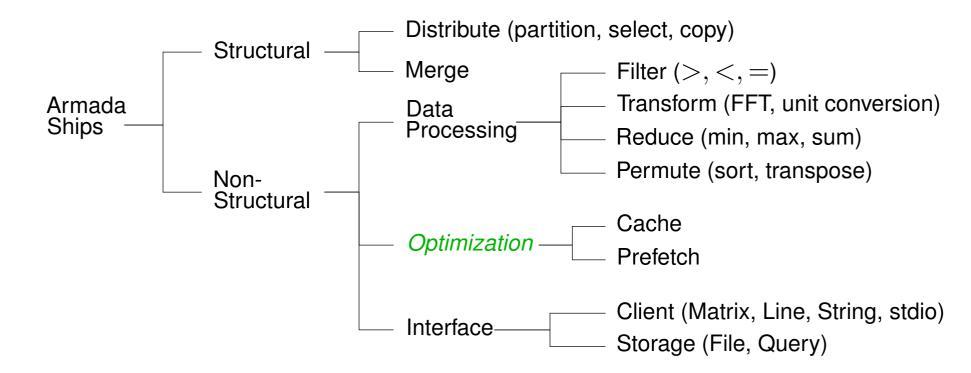
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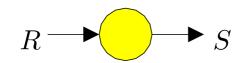
*Data-processing* ships manipulate data, either individually, or in groups as it passes through the ship.



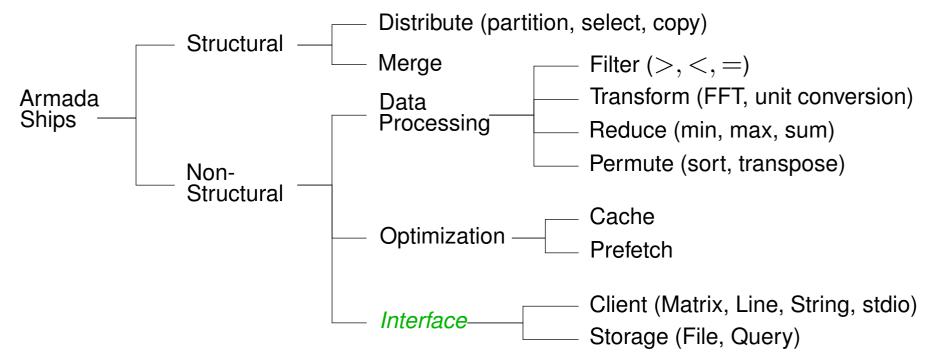
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*Optimization* ships improve I/O performance through latency-reduction techniques like caching and prefetching.

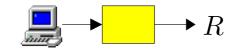


Armada includes a rich set of extensible ship classes.



#### Client-interface ships

convert method calls to a set of requests for data.



Storage-interface ships

access storage devices to process requests.



### **Properties of Ships**

#### Properties of ships are

- used by restructuring and placement algorithms
- assigned by the programmer
- encoded in the ship's definition

#### Properties identify whether a ship

- is data- or request-equivalent
- increases or decreases data flow,
- is parallelizable

A sequence A is *equivalent* to sequence B (denoted  $A \equiv B$ ) if B is a permutation of A, or if B is a set of subsequences that partition A.

#### **Examples:**

$$\begin{cases}
1, 2, 3, 4, 5 \} &\equiv \{2, 3, 5, 1, 4 \} \\
\{1, 2, 3, 4, 5 \} &\equiv \{\{2, 3\}, \{1, 4, 5\} \} \\
\{1, 2, 3, 4, 5 \} &\equiv \{\{2, 3\}, \{1, 5, 4\} \}
\end{cases}$$

In other words, order does not matter.

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A *request-equivalent* ship produces request sequence equivalent to its input.

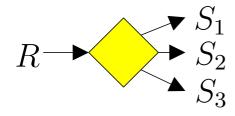
A *data-equivalent* ship produces data sequence equivalent to its input.

Most structural ships are both request and data-equivalent.

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Distribution ships partition requests or data

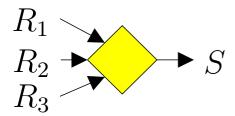
- $S_1$ ,  $S_2$ , and  $S_3$  are subsequences of R.
- $R \equiv \{S_1, S_2, S_3\}$



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Merge ships interleave requests or data

- $R_1$ ,  $R_2$ , and  $R_3$  are subsequences of S.
- $\{R_1, R_2, R_3\} \equiv S$



### **Ships that Change Data Flow**

Data-reducer: a ship that decreases the data flow

- filter
- compress
- reduce (min, max, sum)

Data-increaser: a ship that increases the data flow

- cache
- decompress

### Parallelizable Ships

Parallelizable: a ship that can transform into multiple ships

- process requests and data in parallel
- parallelized by "swapping" with structural ships
- parallel version produces equivalent output

Types of parallelizable ships: replicatable, recursive

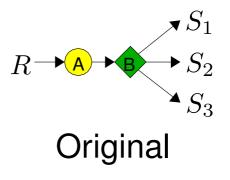
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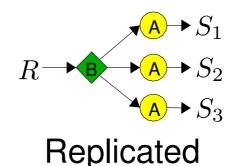
Parallelizable: a ship that can transform into multiple ships

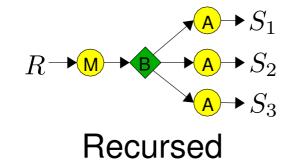
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Types of parallelizable ships: *replicatable*, *recursive* 

#### Right-parallelizable







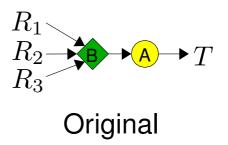
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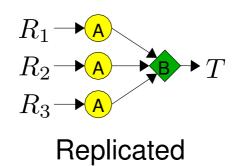
Parallelizable: a ship that can transform into multiple ships

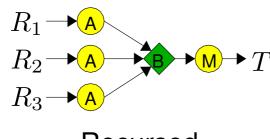
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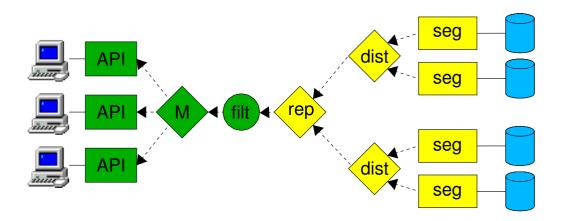
#### Left-parallelizable



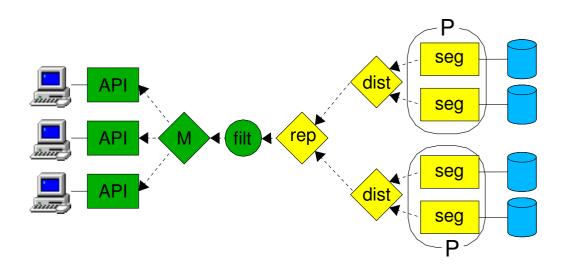


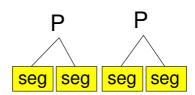


- Syntactically easy to describe (we use XML)
- Easy to manipulate internally

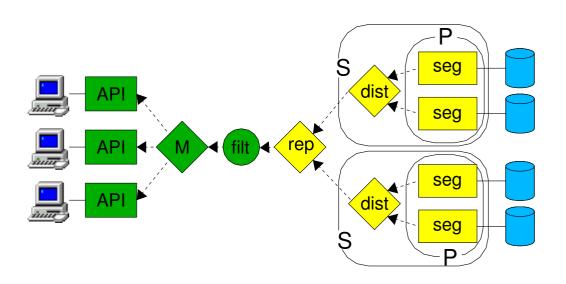


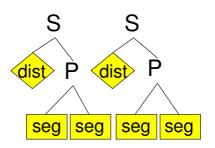
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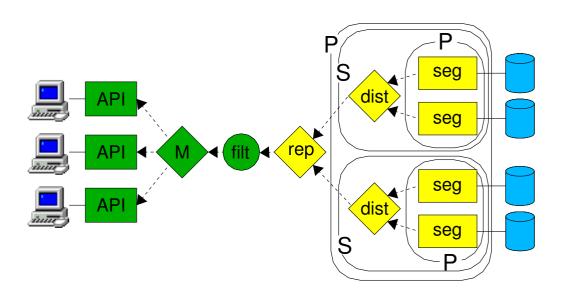


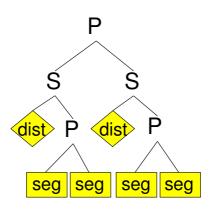
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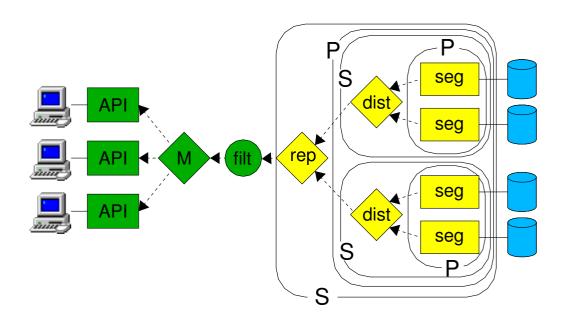


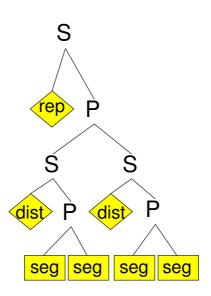
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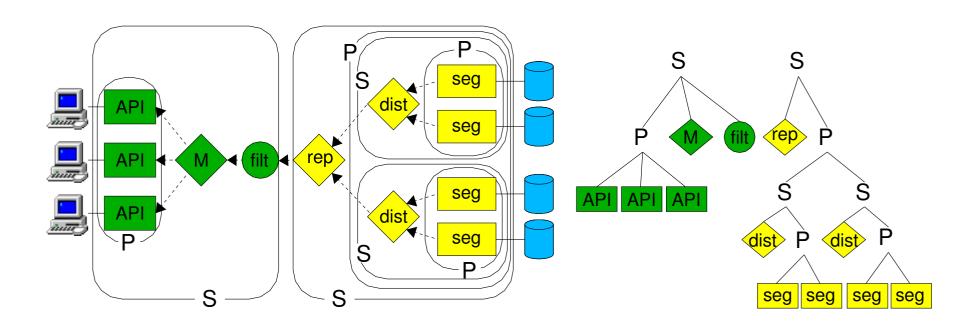


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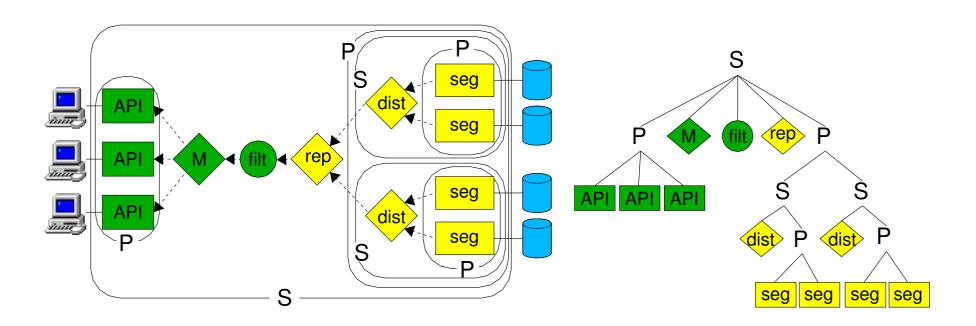




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# **Graph Restructuring**

#### Goals:

- remove bottlenecks (increase parallelism)
- allow effective placement of ships

We restructure by *swapping* adjacent ships in the SP-tree

- increase parallelism by swapping parallelizable ships with structural ships
- reduce network traffic on slow links by
  - moving data-reducing ships toward data source,
  - moving data-increasing ships toward data dest

# The Restruct Algorithm

The RESTRUCT algorithm traverses the SP-tree (depth-first) from node N, revisiting when necessary (all series and parallel nodes are initially marked dirty).

- 1. if N is a leaf or clean (base case)
  - (a) return
- 2. else if N is a parallel node
  - (a) RESTRUCT each child of N
- 3. else if N is a series node
  - (a) create a new series node S
  - (b) while N has children
    - i.  $child \leftarrow$  remove leftmost child of N
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- 4. mark N clean

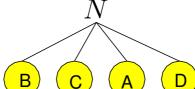
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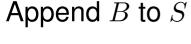
Append B to S

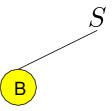


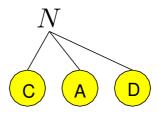
S



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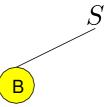


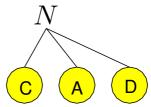




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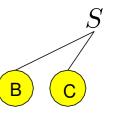


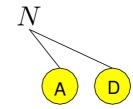




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swap?

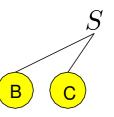
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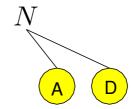




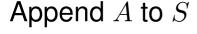
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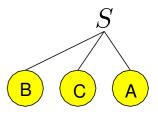


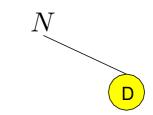




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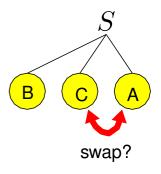


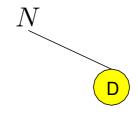




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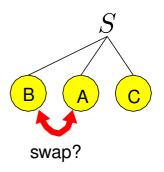


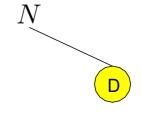




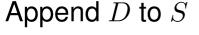
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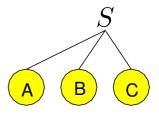


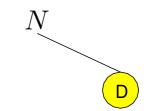




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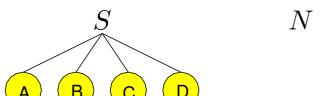






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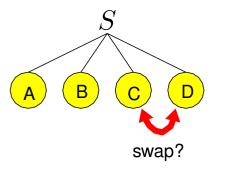


Append D to S

The RESTRUCT algorithm traverses the SP-tree (depth-first) from node N, revisiting when necessary (all series and parallel nodes are initially marked dirty).

- 1. if N is a leaf or clean (base case)
  - (a) return
- 2. else if N is a parallel node
  - (a) RESTRUCT each child of N
- 3. else if N is a series node
  - (a) create a new series node S
  - (b) while N has children
    - i.  $child \leftarrow$  remove leftmost child of N
    - ii. append child to S
    - iii. SLIDE child left
  - (c)  $N \leftarrow S$
- 4. mark N clean



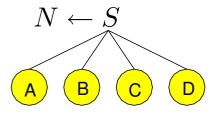


N

The RESTRUCT algorithm traverses the SP-tree (depth-first) from node N, revisiting when necessary (all series and parallel nodes are initially marked dirty).

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    - ii. append child to S
    - iii. SLIDE child left
  - (c)  $N \leftarrow S$
- 4. mark N clean

Assign S to N



Conditions for swapping two series-connected ships (labeled A and B)

- A and B are commutative (A or B is request-equivalent and A or B is data-equivalent)
- swapping A and B is beneficial to the application (see next slide), and
- the graph resulting from a swap is an SP-DAG (we allow four configurations).

Conditions for swapping two series-connected ships (labeled A and B)

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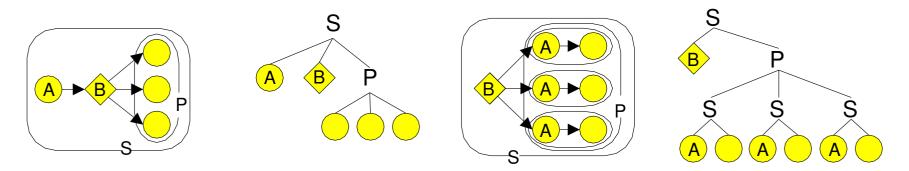
(A) Non-structural, (B) Non-structural



Conditions for swapping two series-connected ships (labeled A and B)

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- the graph resulting from a swap is an SP-DAG (we allow four configurations).

#### (A) Non-structural, (B) Distribution, Parallel node

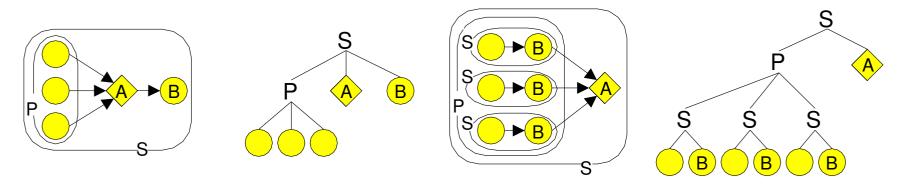


PARALLELIZE right

Conditions for swapping two series-connected ships (labeled A and B)

- A and B are commutative (A or B is request-equivalent and A or B is data-equivalent)
- swapping A and B is beneficial to the application (see next slide), and
- the graph resulting from a swap is an SP-DAG (we allow four configurations).

#### Parallel node, (A) Merge, (B) Non-structural

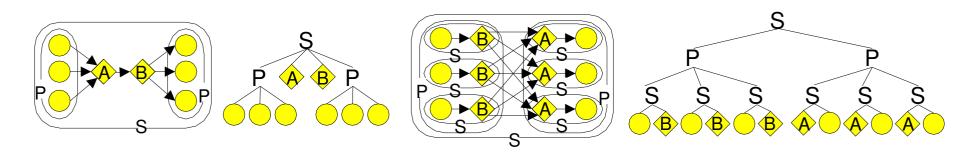


Parallelize left

Conditions for swapping two series-connected ships (labeled A and B)

- A and B are commutative (A or B is request-equivalent and A or B is data-equivalent)
- swapping A and B is *beneficial* to the application (see next slide), and
- the graph resulting from a swap is an SP-DAG (we allow four configurations).

Parallel node, (A) Merge, (B) Distrib, Parallel node



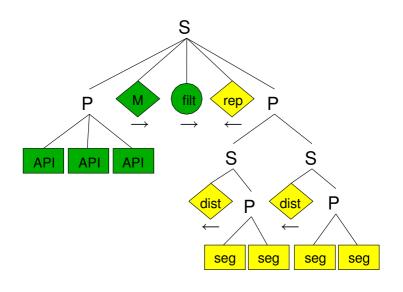
Parallelize right and left

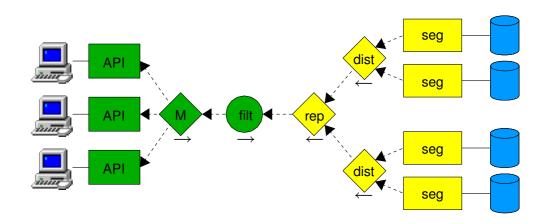
#### **Beneficial Swap**

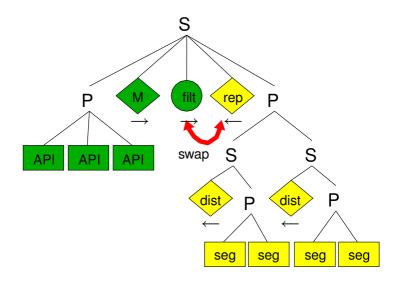
A swap is deemed *beneficial* if it increases parallelism, moves a data-reducing ship closer to the data source, or moves a data-increasing ship closer to data destination.

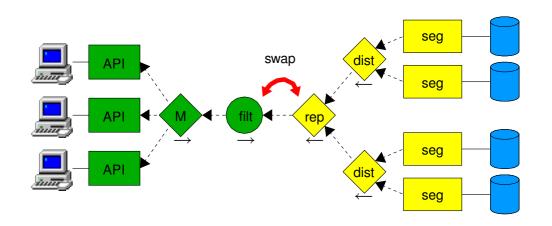
Algorithm to decide a beneficial swap of adjacent ships A and B

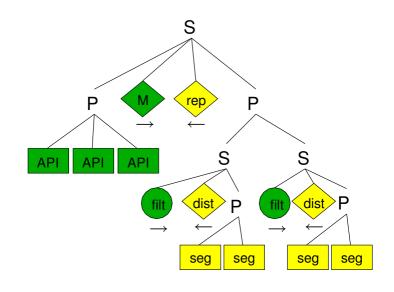
- 1. Assign a preferred direction to each ship (1 for right, -1 for left, or 0)
  - Merge ships prefer to go right (increase parallelism)
  - Distribution ships prefer to go left (increase parallelism)
  - Data-reducing ships prefer to swap toward the data source
  - Data-increasing ships prefer to swap toward the data destination
- 2. return true if preferred direction of A is greater than preferred direction of B
- 3. else return *false*

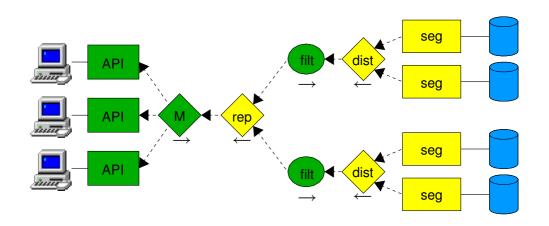


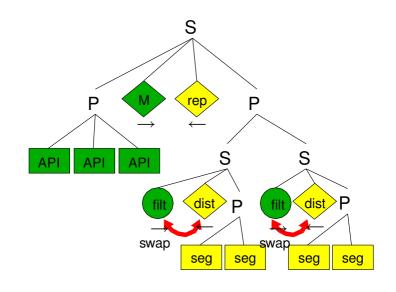


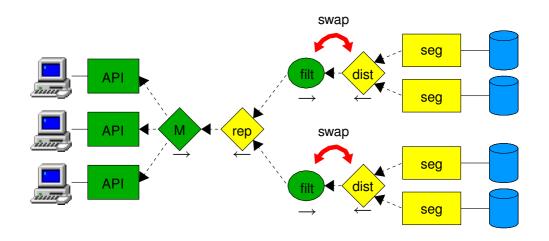


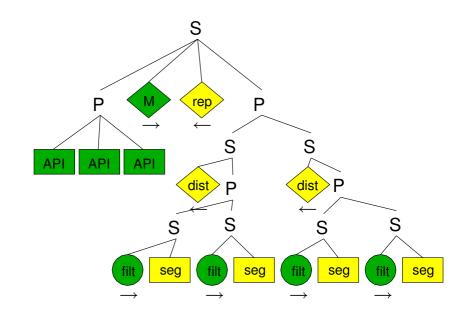


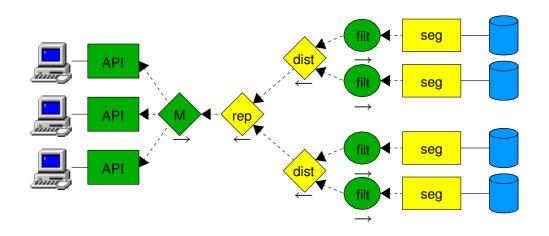


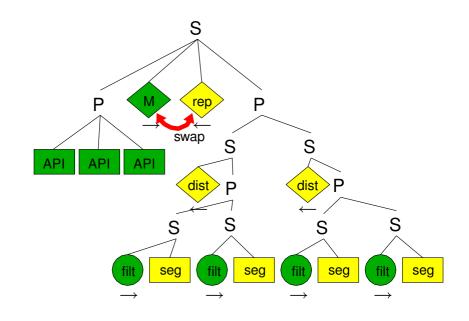


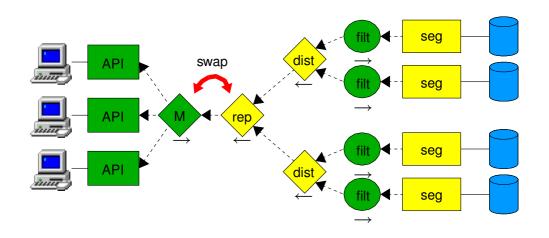


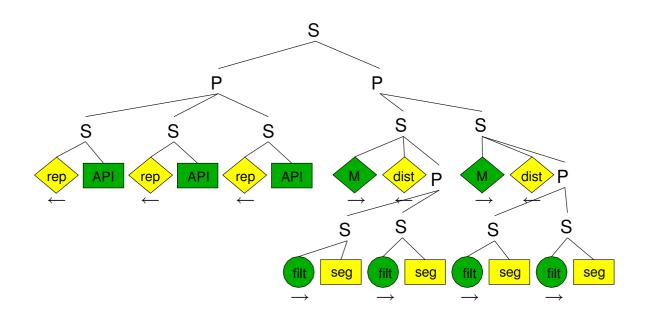


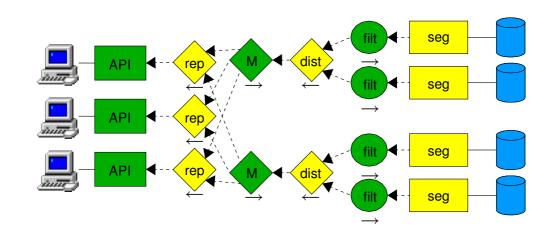












#### **Placement**

#### Hierarchical graph partitioning

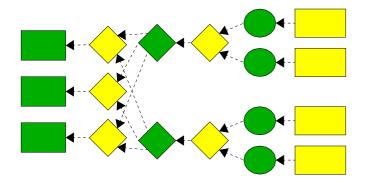
- 1. Partition the ships into k sets (each set represents an administrative domain).
- 2. Partition the ships within each domain to processors provided by domain-level schedulers.

#### The Graph Partitioning Problem

Given graph G(V, E) with weighted vertices and weighted edges, partition the vertices into k sets in such a way to balance the sum of the vertices and to minimize the weights of the edge crossings between sets (NP-hard [Garey et al., 1976]).

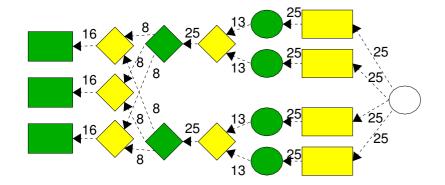
Chaco Graph Partitioning Software [Hendrickson and Leland, SNL]

- 1. Construct model from SP-tree
  - (a) Assign edge weights
  - (b) Assign vertex weights
- 2. partition graph (using CHACO)
- 3. for each domain
  - (a) request procs from domain
  - (b) partition sub-graph



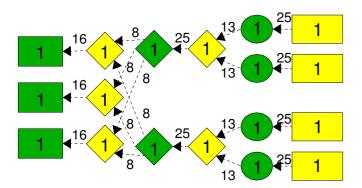
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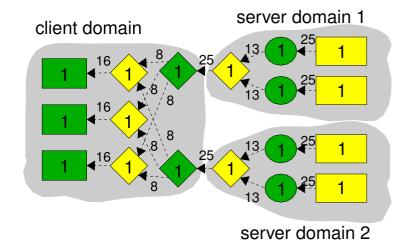
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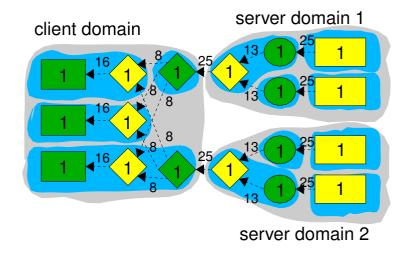
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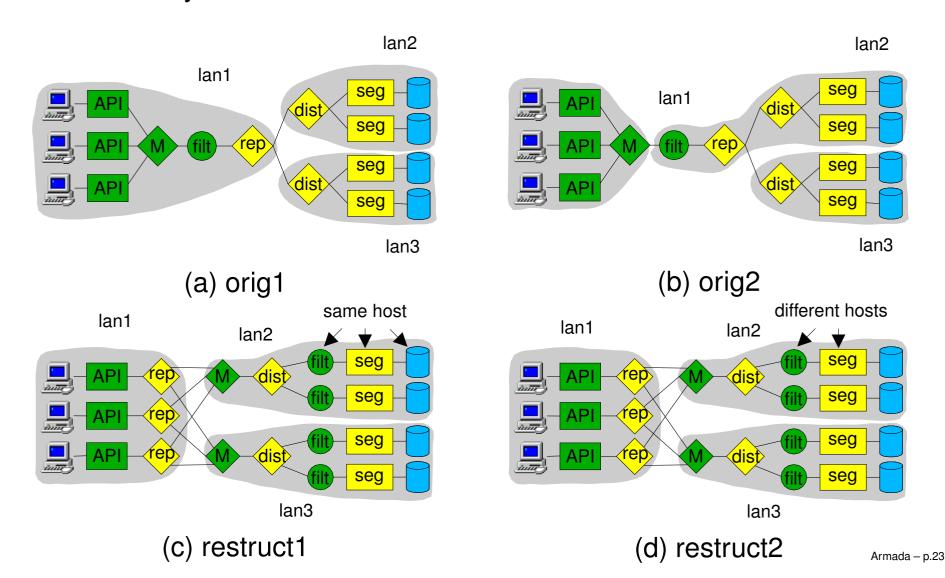
#### **Experiments**

Evaluate performance benefit of restructuring and placement

- Representative application
  - Placement considerations
- File copy and permutation
  - Third-party transfers
  - Data permutations
  - Number of processors required by Armada
- Seismic processing
  - C++ interface
  - Recursive filter
  - Latency effects

#### Representative Application

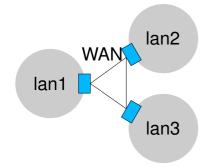
Examined four configurations of the example application with a filter that removed exactly 50% of the data.



# **Experiment Setup**

The area between the blobs represents the WAN

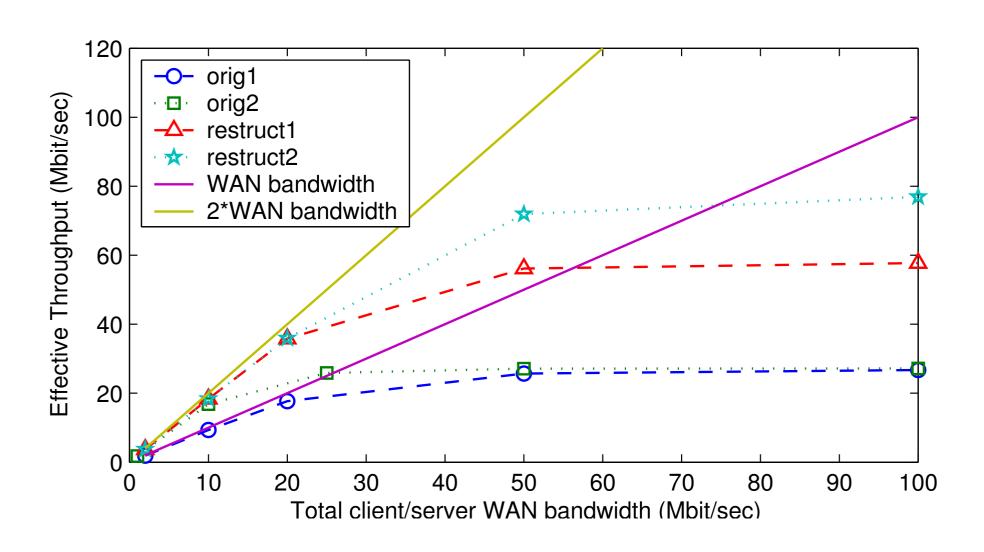
- each LAN connected to the WAN by single router
- each WAN link has limited capacity

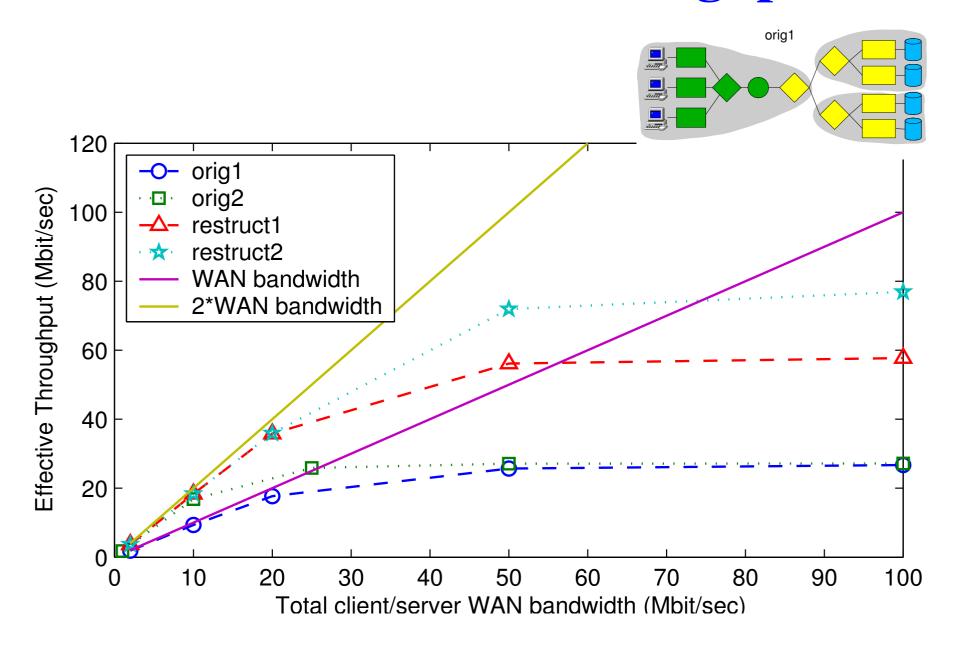


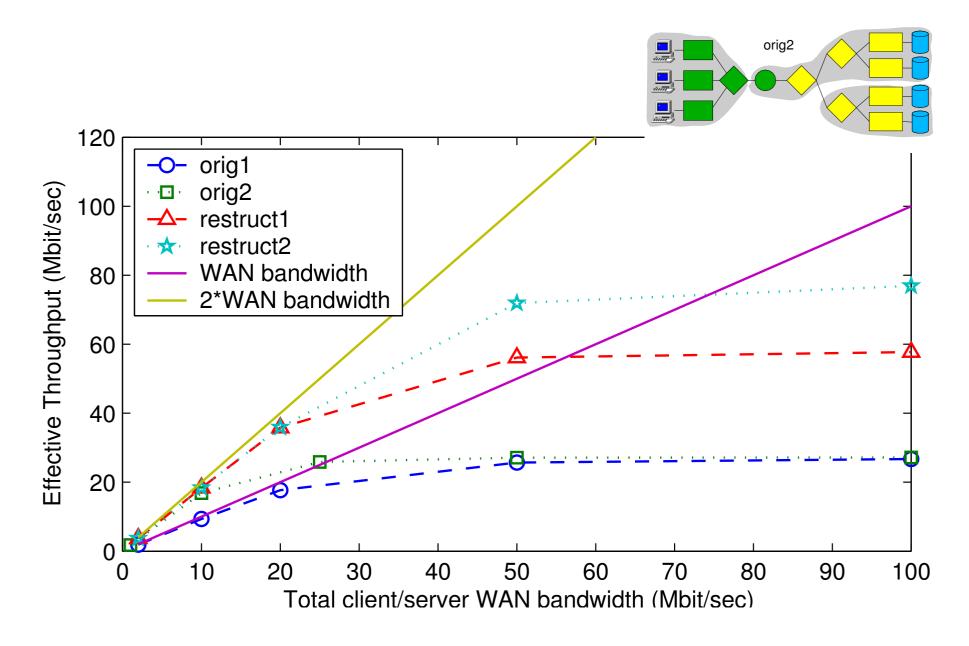
Ran experiments on the Emulab Network Testbed

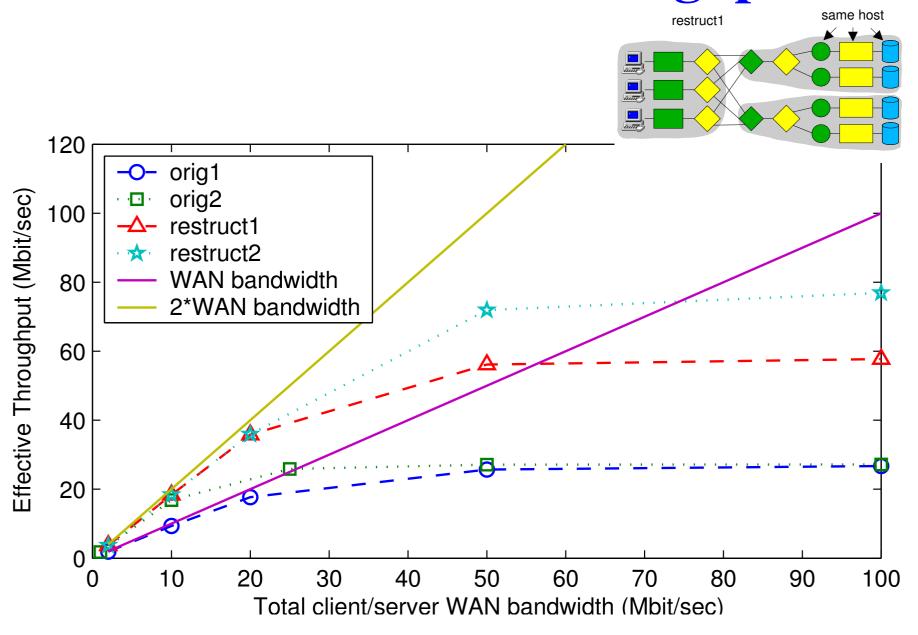
- Three LANs, each with...
  - Five 850 MHz Pentium III processors
  - 100 Mbps switched network (0.15 msec latency)
- WAN consisted of...
  - Three network links with 2.0 msec latency
  - Bandwidth ranged from 2 to 100 Mbps

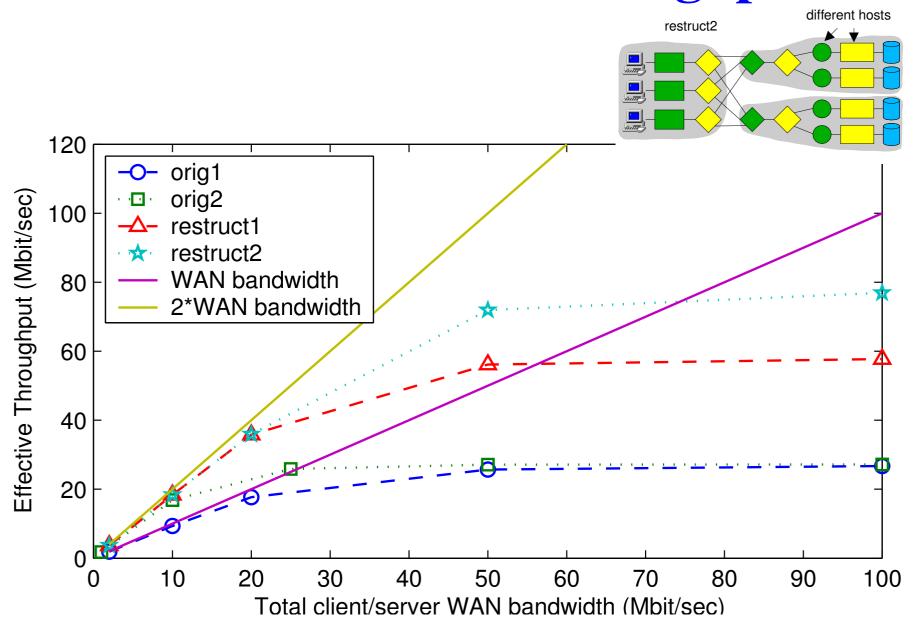
#### **Results: Effective Throughput**

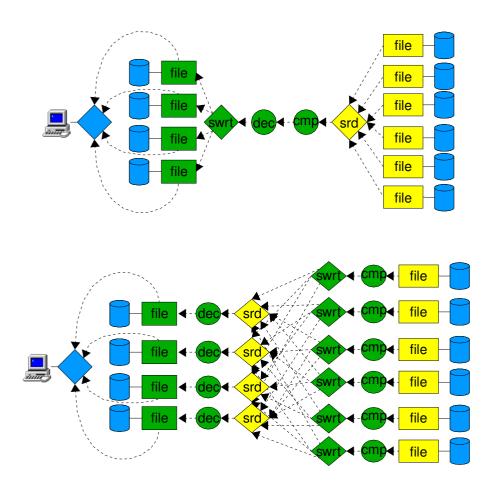


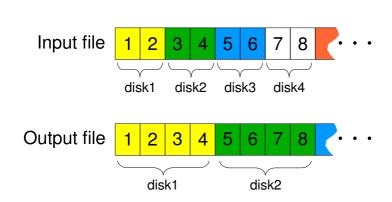


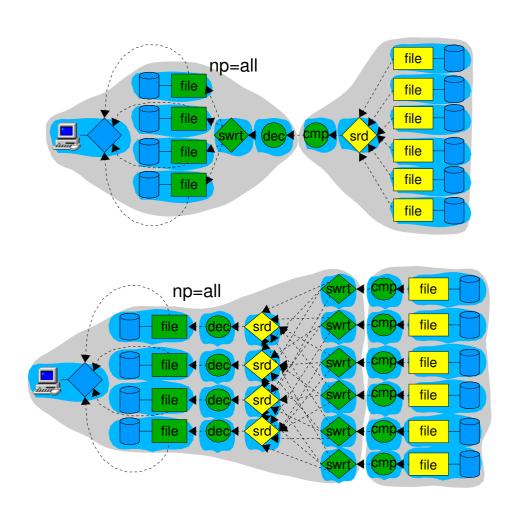


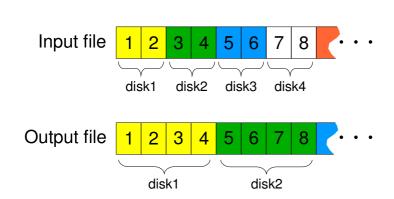


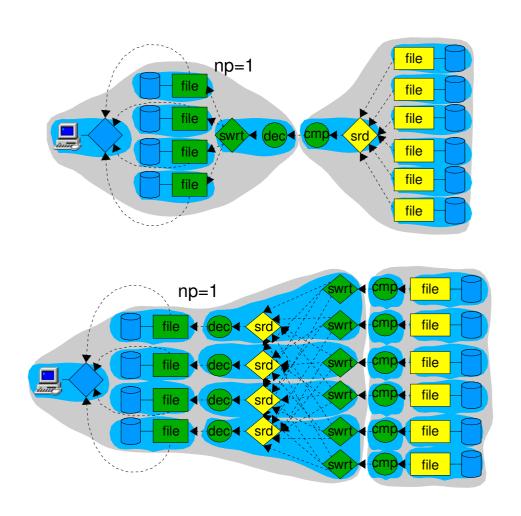


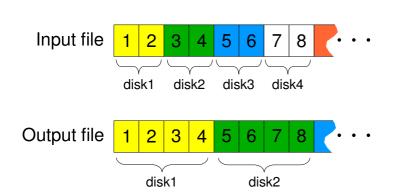


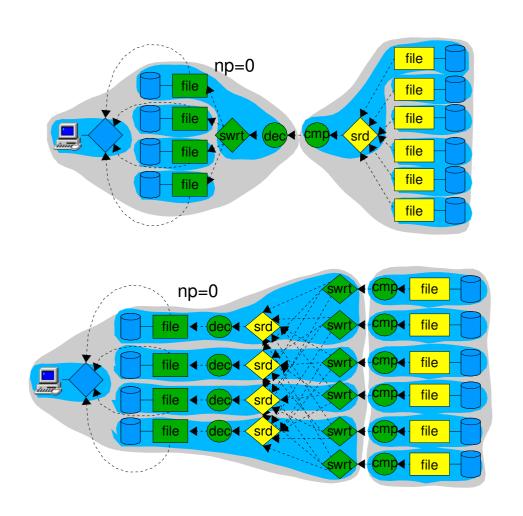


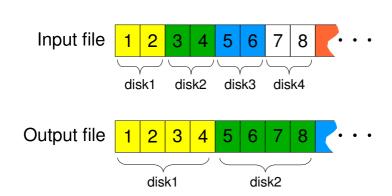




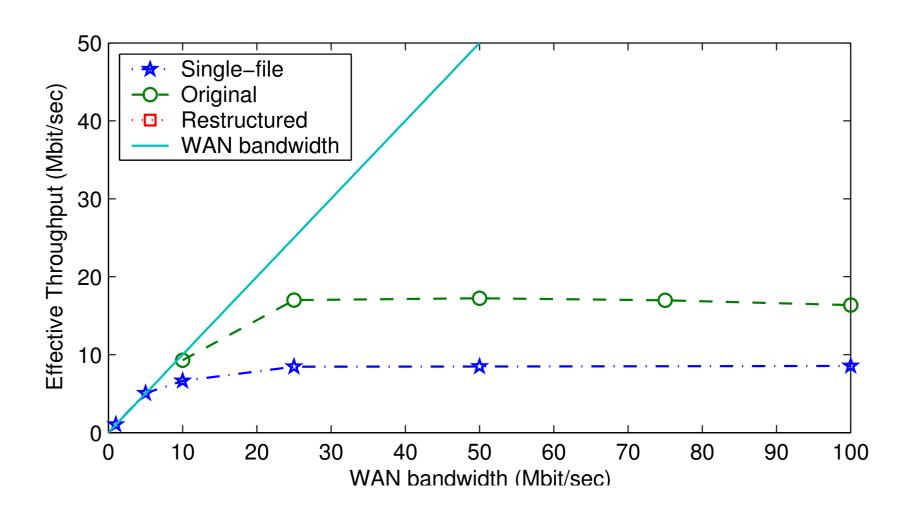




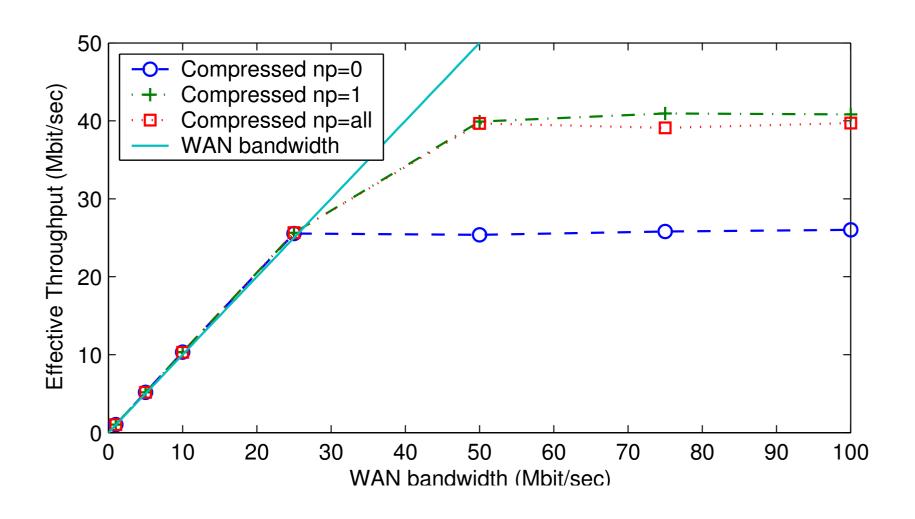




### Results (effective throughput)



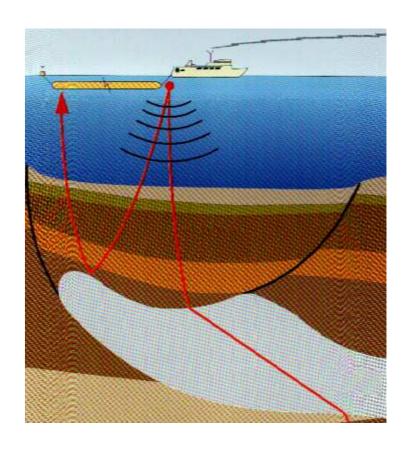
### Results (different placements)

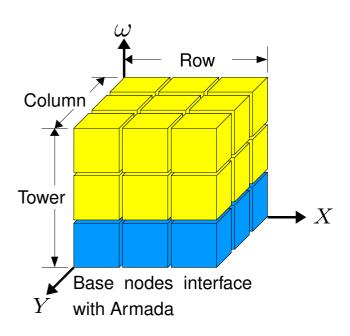


#### **Post-Stack Seismic Imaging**

#### Properties of seismic processing

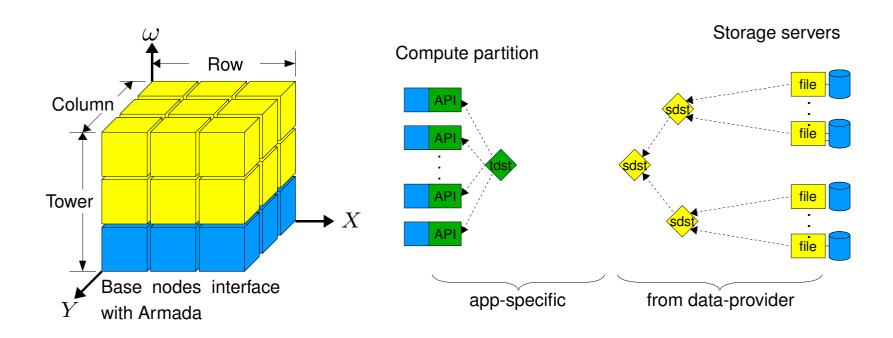
- Compute intensive
- Large (terabyte) data sets
  - Collections of files (> 1K)
  - Each file contains a set of *traces* (recorded pressure waves)
- Preprocessing
  - Stack co-located traces
  - FFT time traces
  - Distribute frequencies to compute nodes





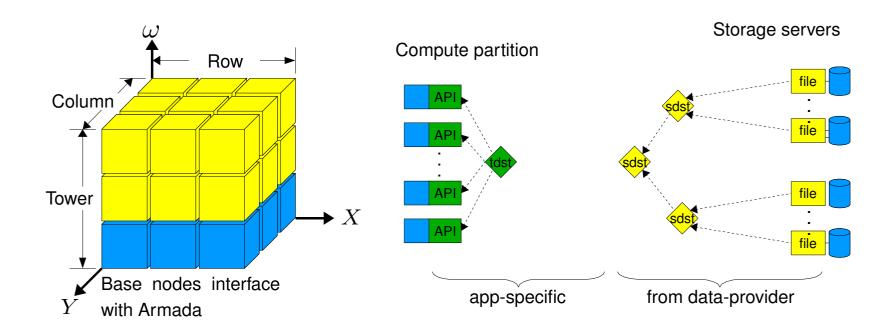
Connect with the data provider and describe compute node distribution

```
// called by all nodes,
// ... node0 gets graph from data provider
// ... constructor decomposes data (3D block decomposition)
TraceDataset dataset(comm, pmesh, providerURL)
```



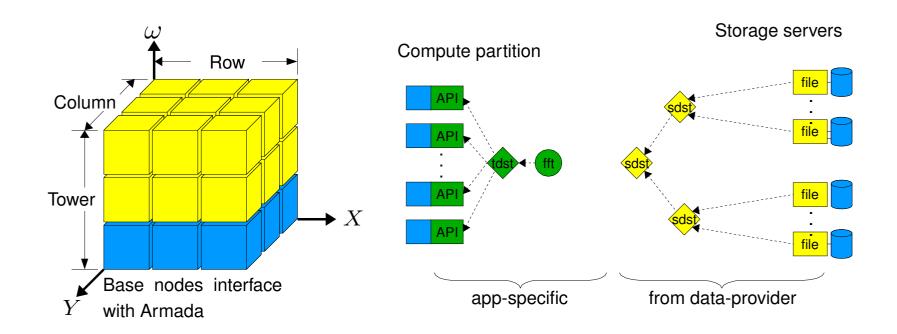
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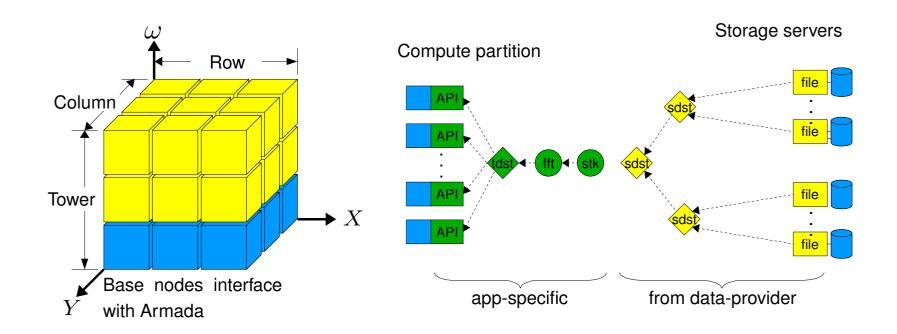
#### Append operators.

```
// called by node0
dataset.appendOp(new FFTOp());
```



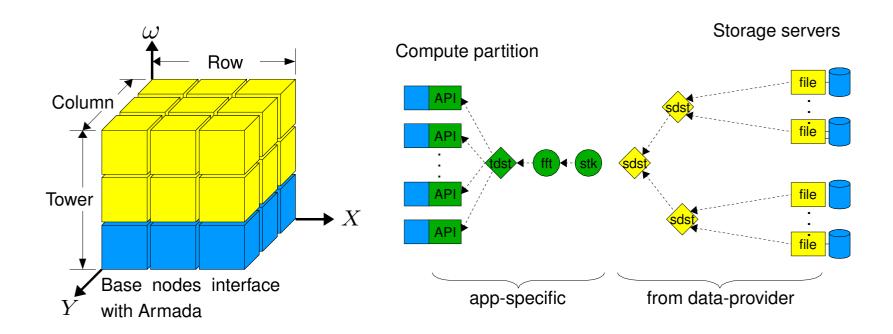
#### Append operators.

```
// called by node0
dataset.appendOp(new FFTOp());
dataset.appendOp(new StackOp());
```

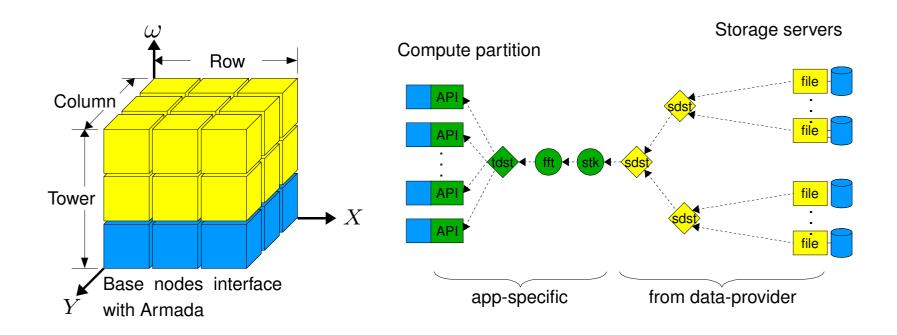


#### Append operators.

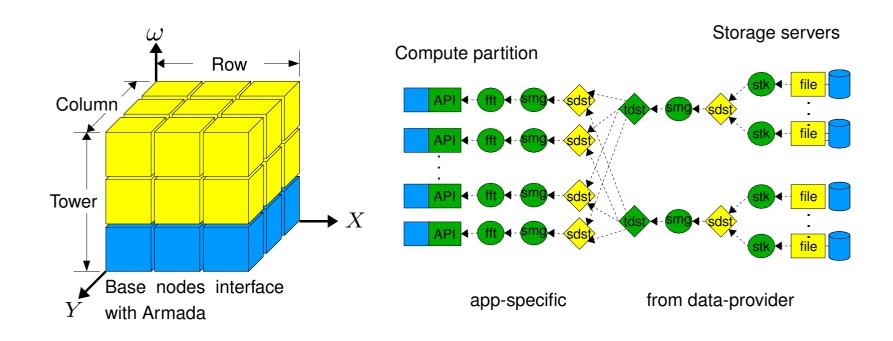
```
// called by node0
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dataset.appendOp(new StackOp());
```



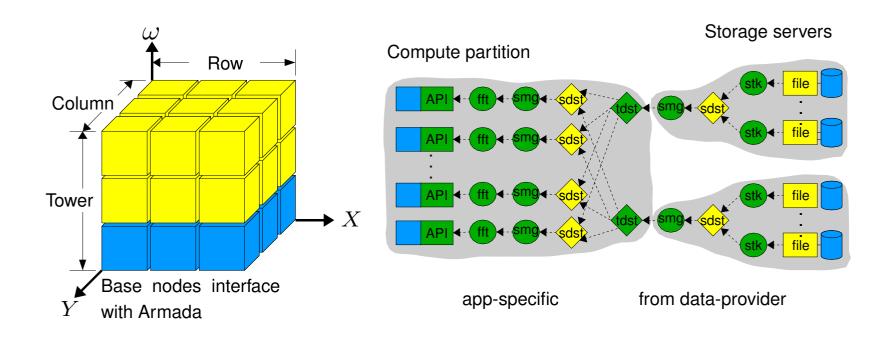
```
// called by node0
dataset.open();
```



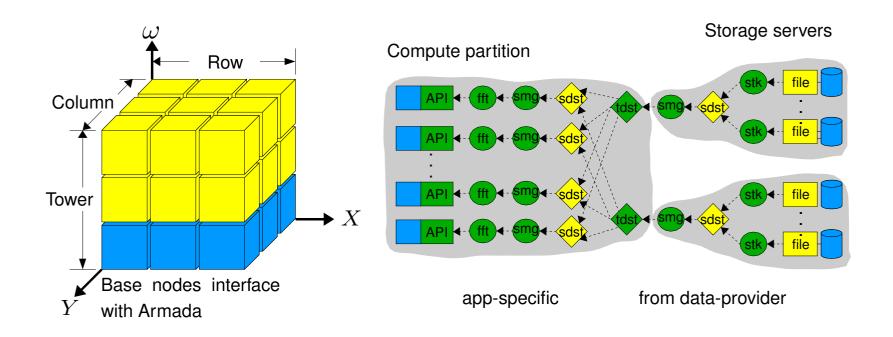
```
// called by node0
dataset.open();
// ... connect app-specific with data-provider portion
```



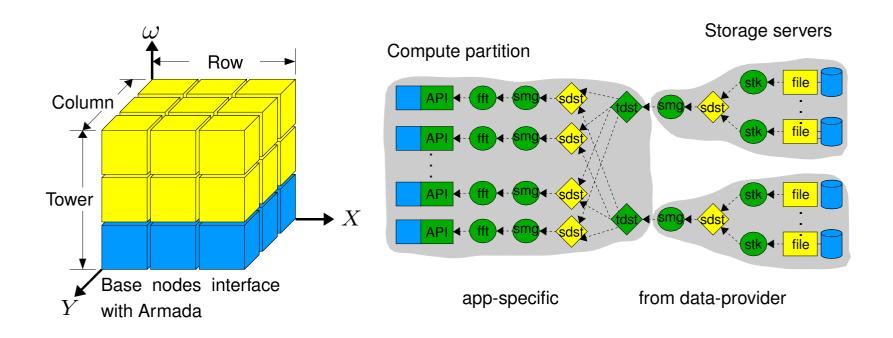
```
// called by node0
dataset.open();
// ... connect app-specific with data-provider portion
// ... restructure graph
```



```
// called by node0
dataset.open();
// ... construct entire Armada graph
// ... restructure graph
// ... assign placement
```



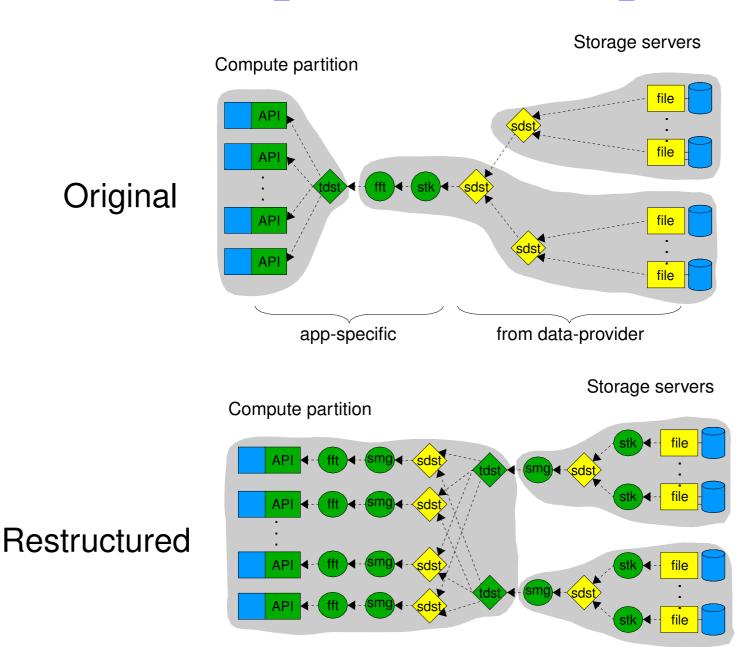
```
// called by node0
dataset.open();
// ... construct entire Armada graph
// ... restructure graph
// ... assign placement
// ... deploy
```



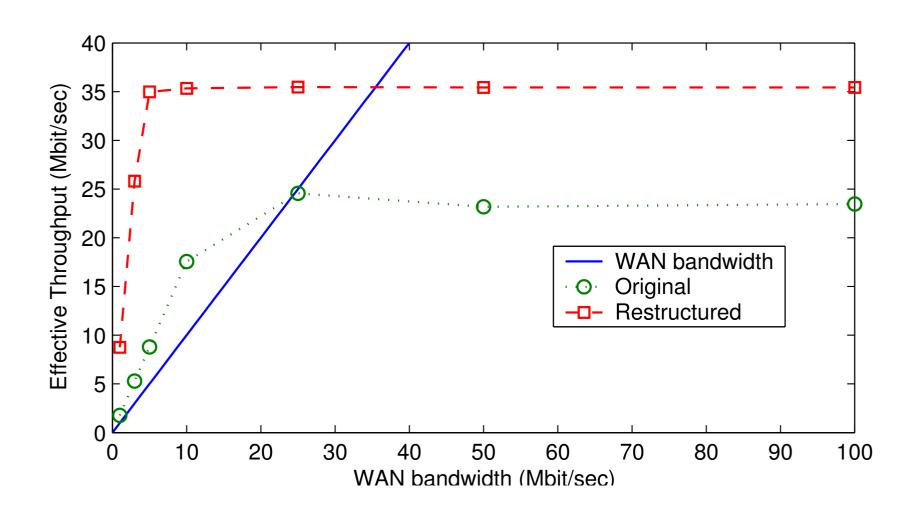
#### Collectively read dataset.

```
// called by all procs
int size=dataset.getLocalSize();
float *data = new float[size];
dataset.read(data);
// do computation ...
```

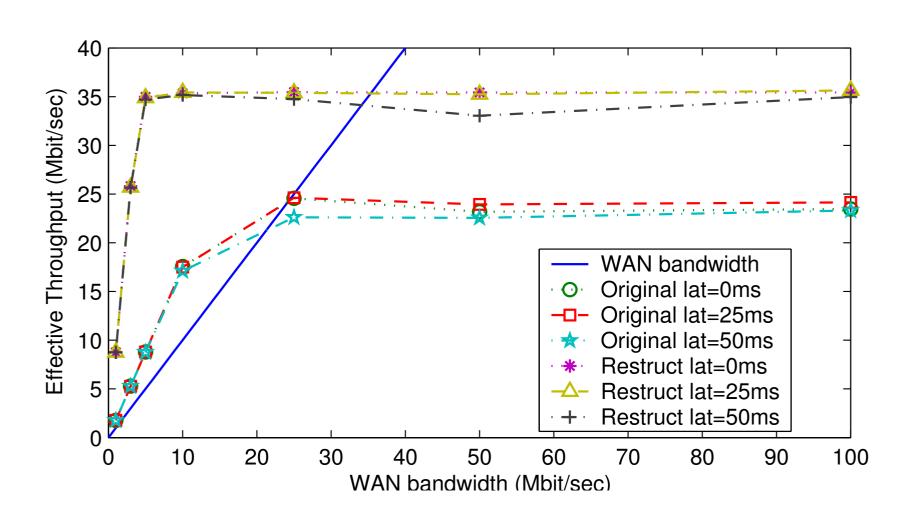
### **Experiment Setup**



#### Results (effective throughput)



#### Results (different latencies)



#### **Related Work**

Parallel processing of I/O streams

- PS<sup>2</sup>[Messerli, 1999]
  - data-flow model with automatic parallelization
- DataCutter [Spencer et al., 2002]
  - component-based, analytic model to decide parallelization

Armada does not force the whole application into a data-flow model Armada widens data flow for parallel clients and parallel servers

Operation re-ordering to improve data flow, e.g., in databases

- dQUOB [plale et al. 2000]
  - optimize query tree to move high-filtering portions close to data
  - exploit well-defined properties associated with query processing

Armada provides a more general approach

#### **Future Work**

#### Other Applications

- fMRI application (time-series analysis of brain data)
- Can components be reused between applications?

#### Modifications to BENEFICIAL and COMMUTATIVE

- Non-greedy methods
- Analytic models to approximate benefit

#### **Placement**

- incorporate domain-specific information into the partitioner (compute capacity, memory capacity, etc...)
- dynamic re-deployment when network conditions change

Tuning for cluster computing (in addition to the grid)

#### Summary

#### The Armada framework

- data provider can describe complex distributed data sets
- application describes processing required before computation
- data-flow model provides a "latency-tolerant" approach

#### Restructuring algorithm

- arranges graph to provide end-to-end parallel I/O
- enables effective placement of data-processing components

#### **Placement**

- domain assignments to minimize data flow.
- host assignments based on administrative domain policies.

Experiments demonstrate good performance in multiple environments.

# Efficient I/O for Computational Grid Applications

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http://www.cs.dartmouth.edu/~dfk/armada/

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