

# Cluster Computing with DryadLINQ

Mihai Budiu

Microsoft Research Silicon Valley

**IEEE Cloud Computing Conference**

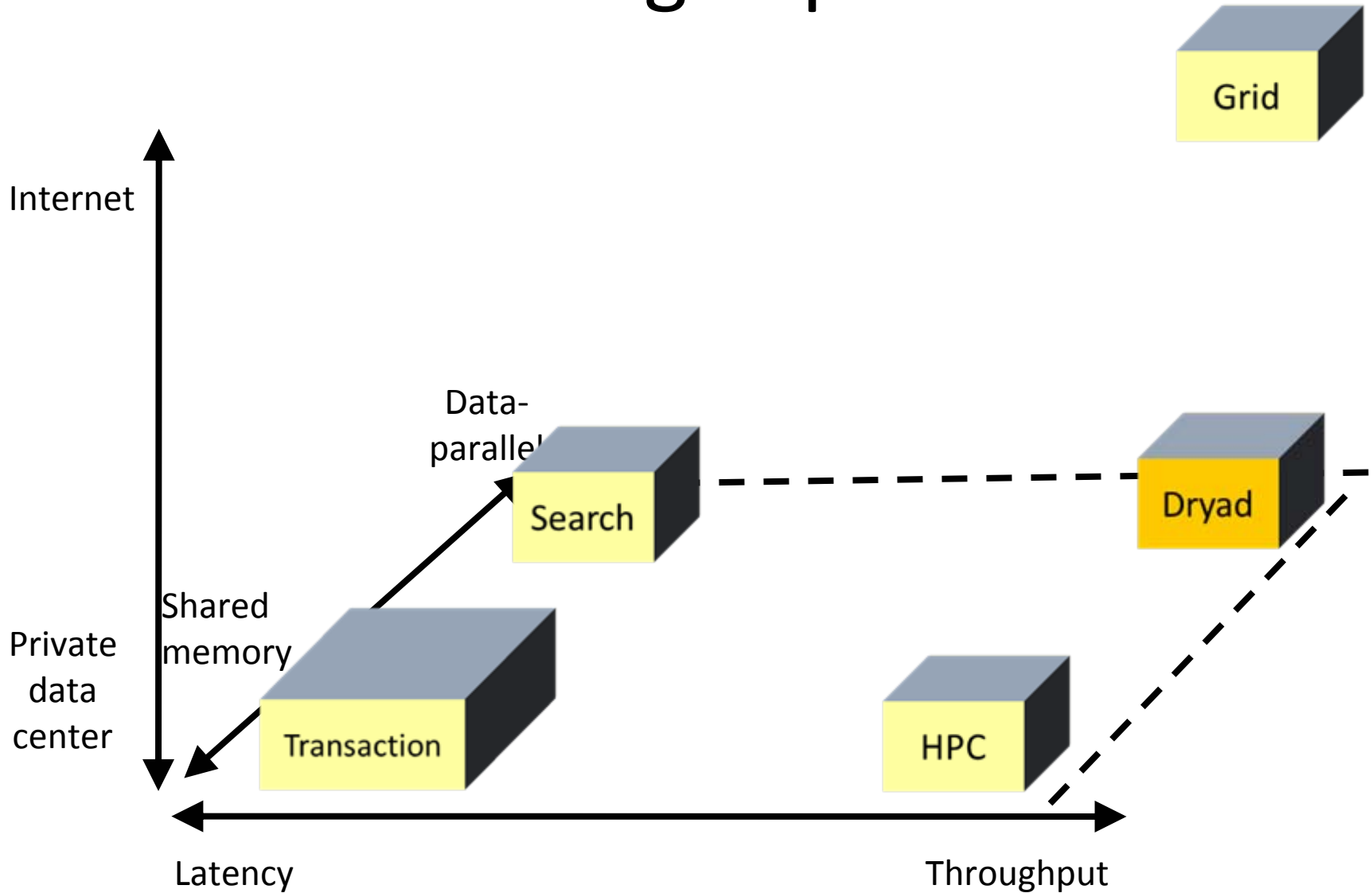
July 19, 2008



# Goal



# Design Space

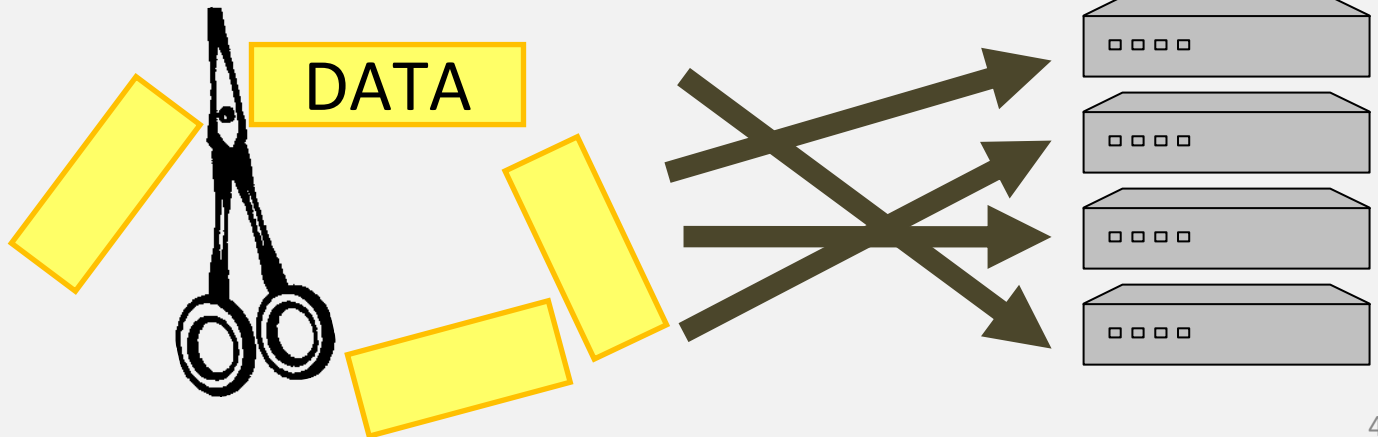
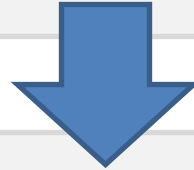


# Data Partitioning

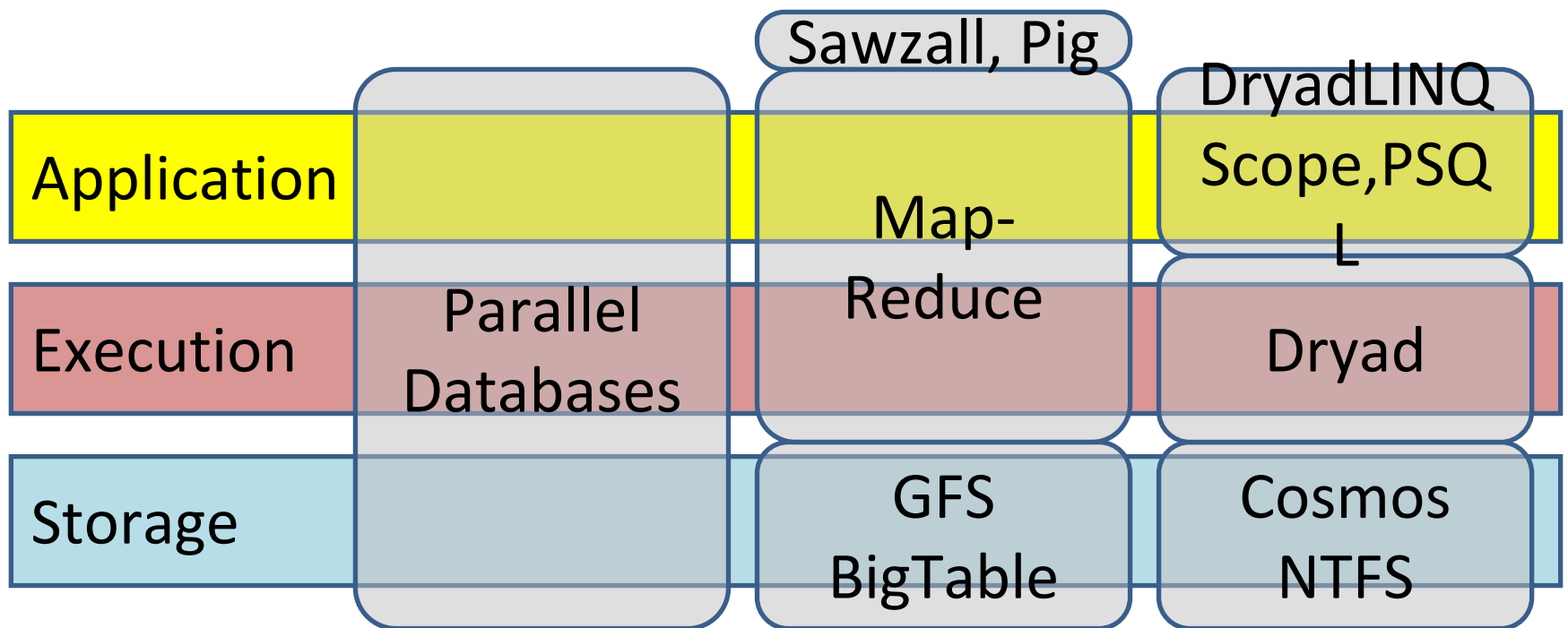
DATA



RAM



# Data-Parallel Computation



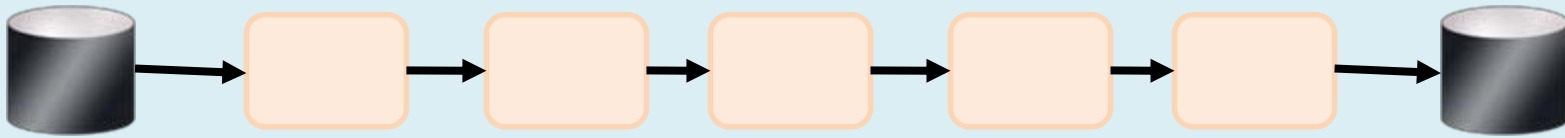
# Outline

- Introduction
- Dryad
- DryadLINQ
- Applications

# 2-D Piping

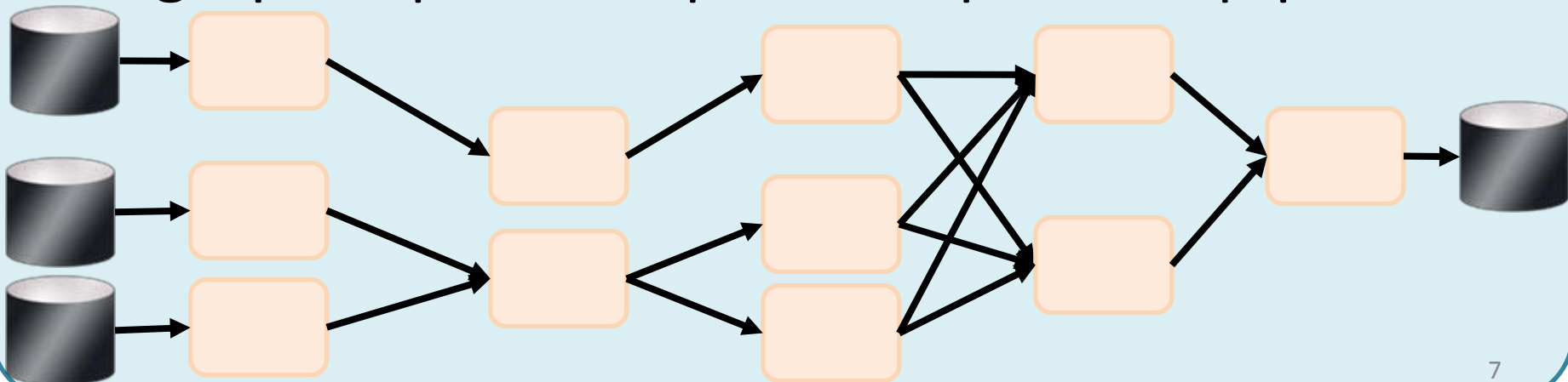
- Unix Pipes: 1-D

grep | sed | sort | awk | perl

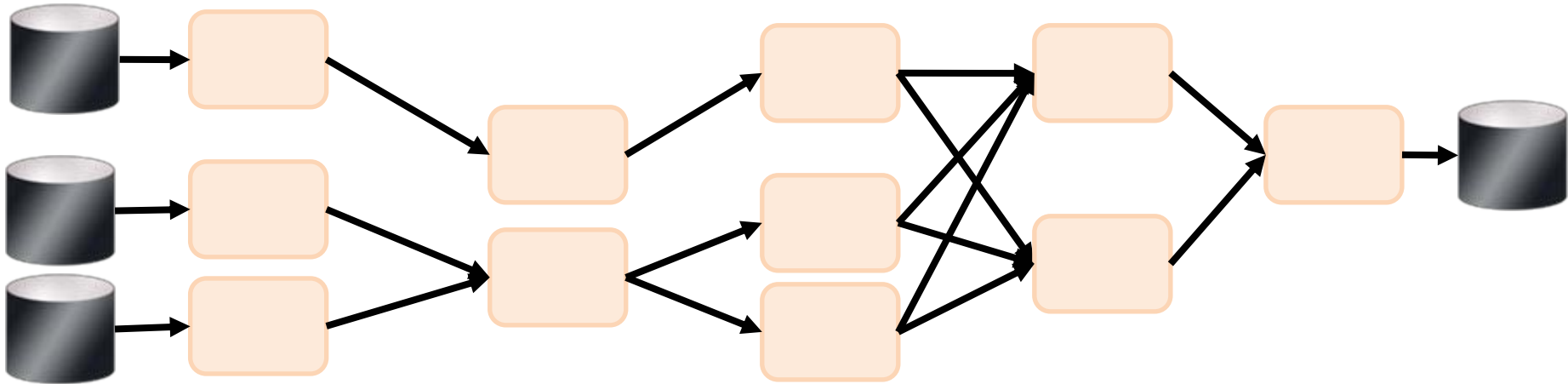


- Dryad: 2-D

grep<sup>1000</sup> | sed<sup>500</sup> | sort<sup>1000</sup> | awk<sup>500</sup> | perl<sup>50</sup>

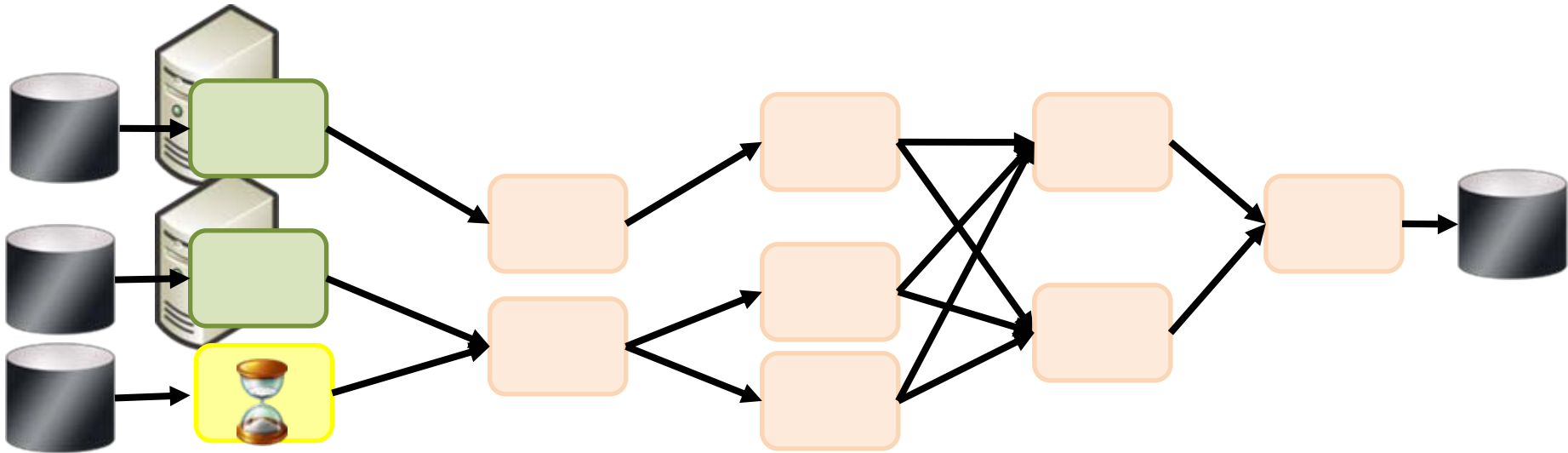


# Virtualized 2-D Pipelines

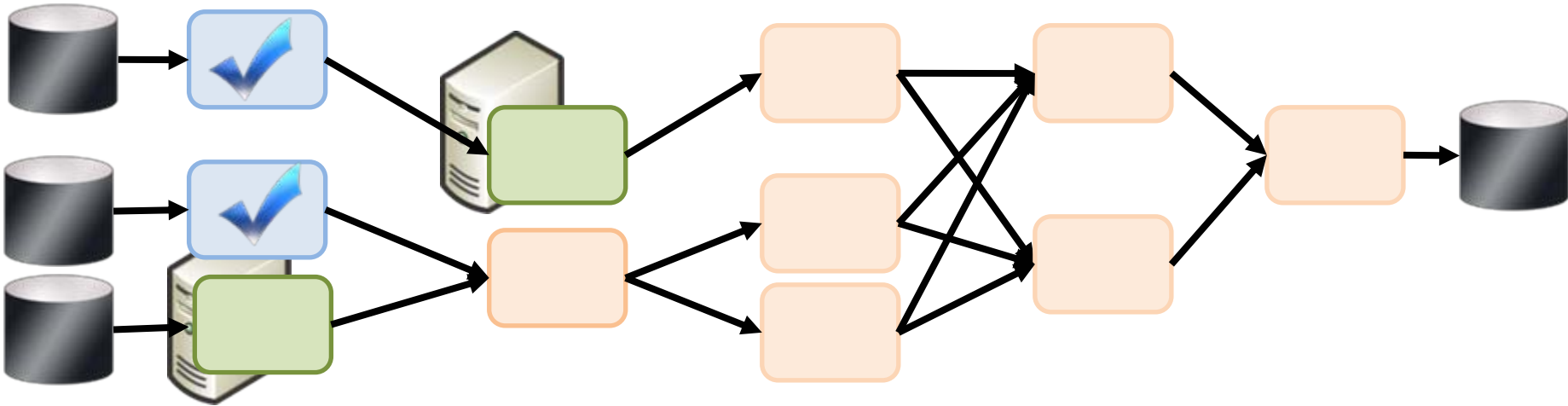




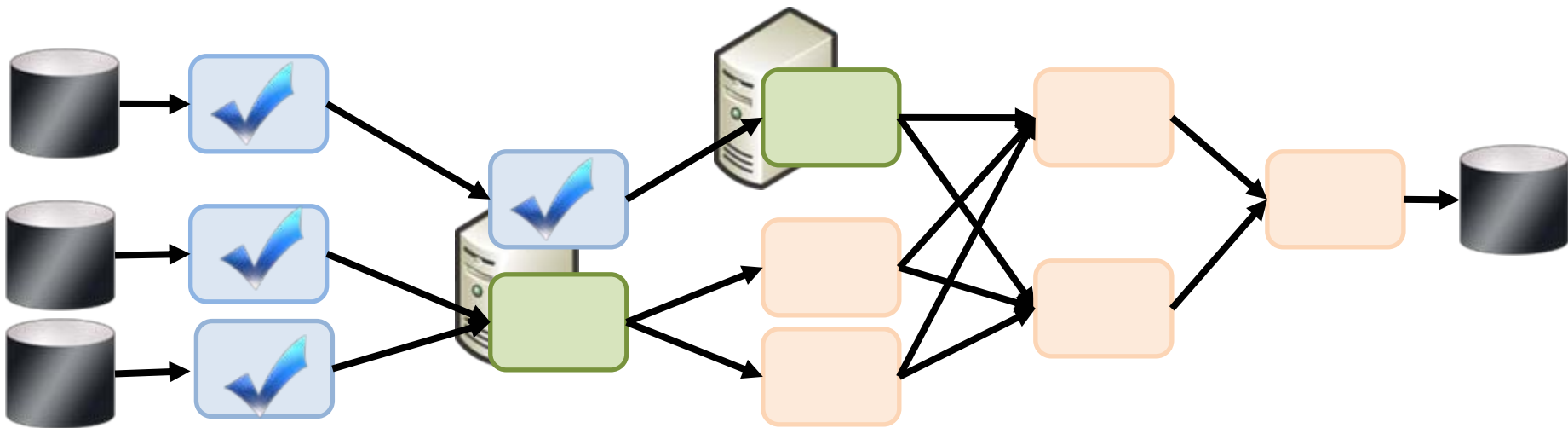
# Virtualized 2-D Pipelines



# Virtualized 2-D Pipelines

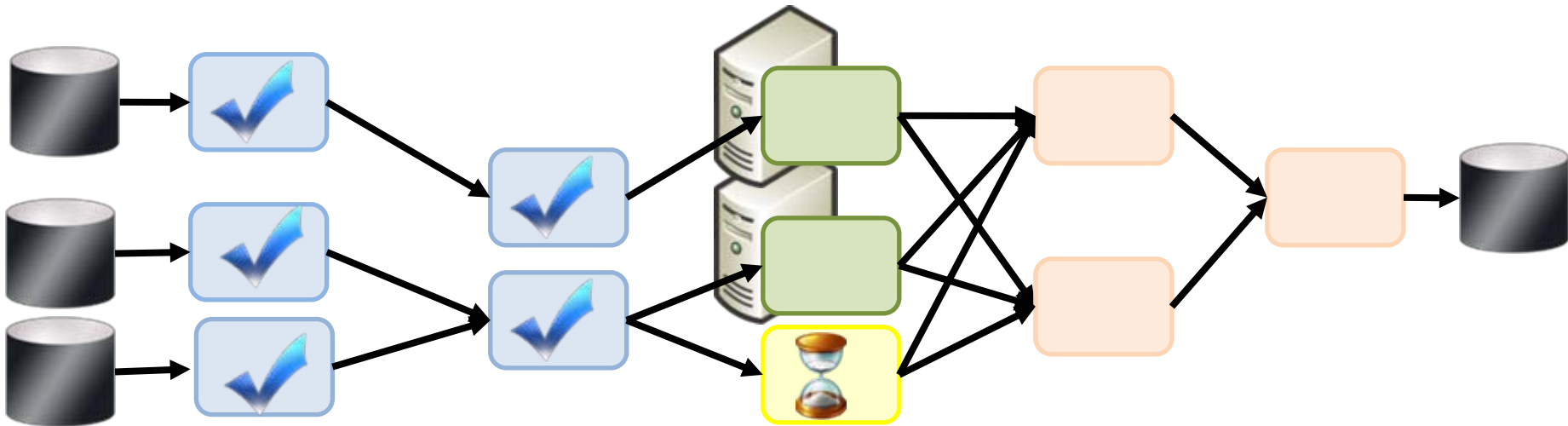


# Virtualized 2-D Pipelines

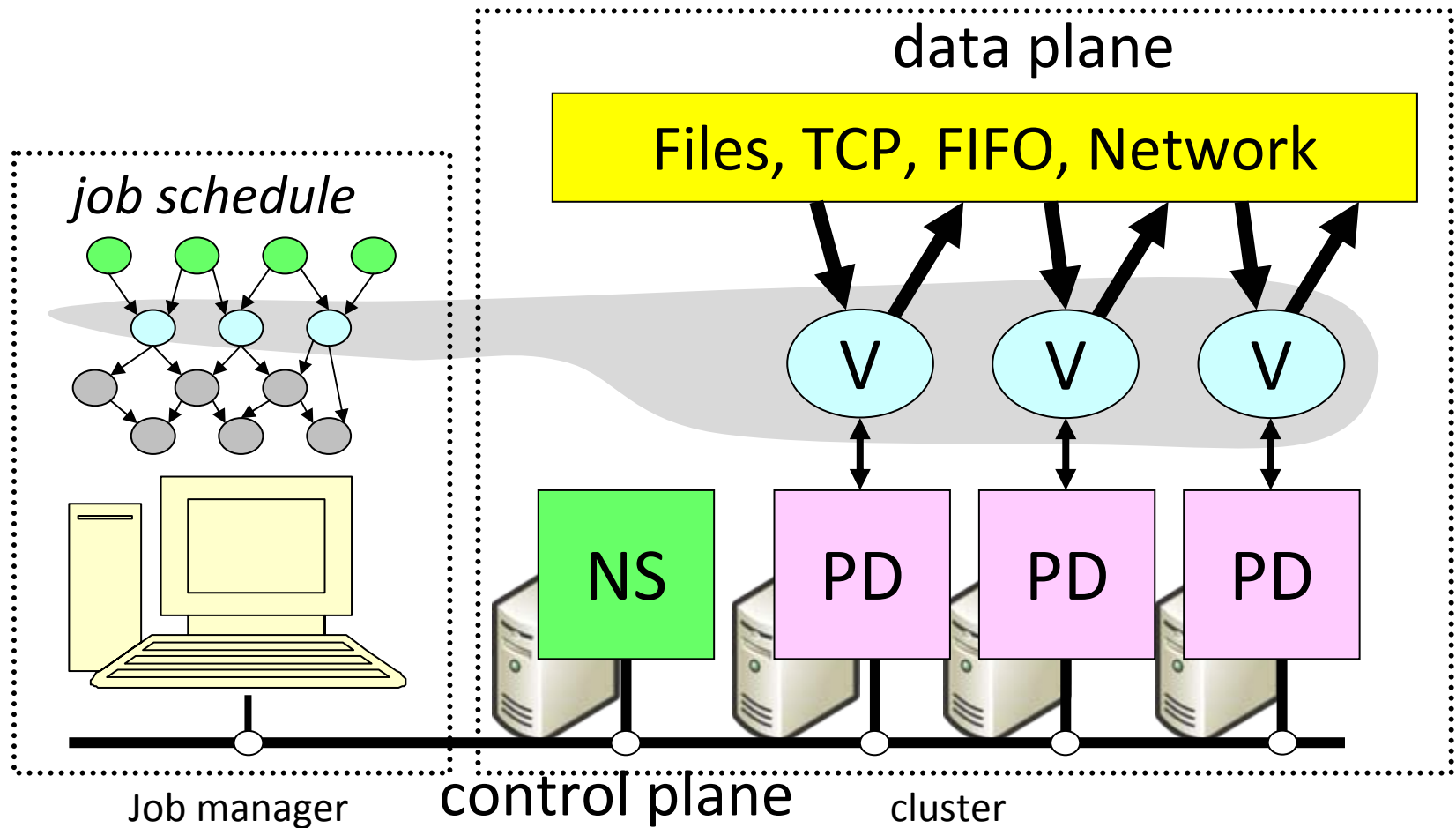


# Virtualized 2-D Pipelines

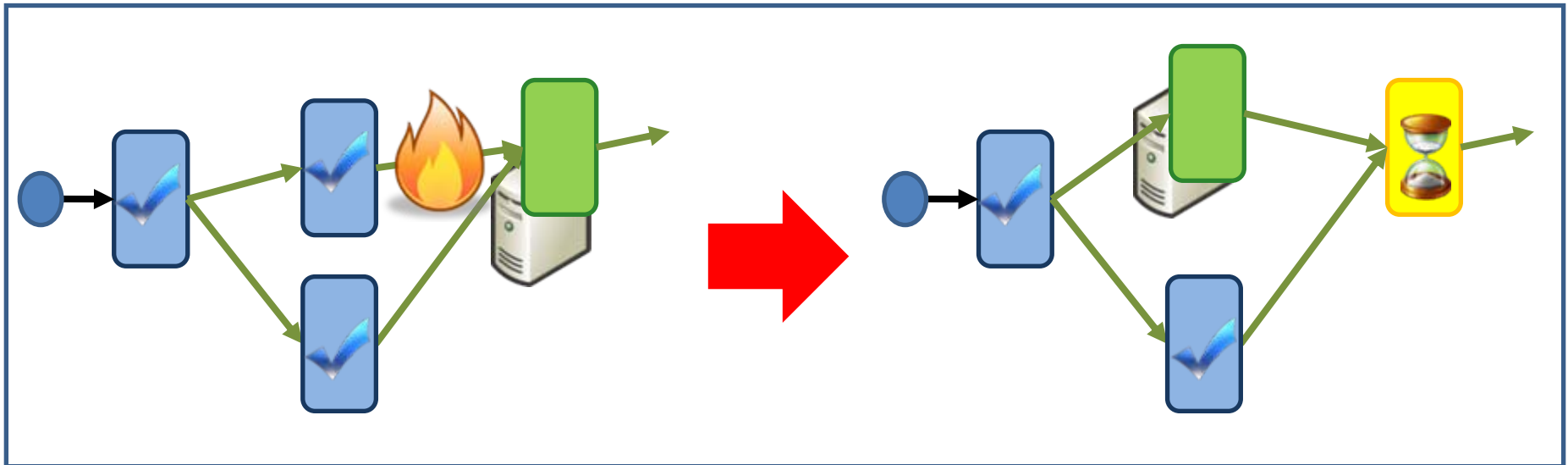
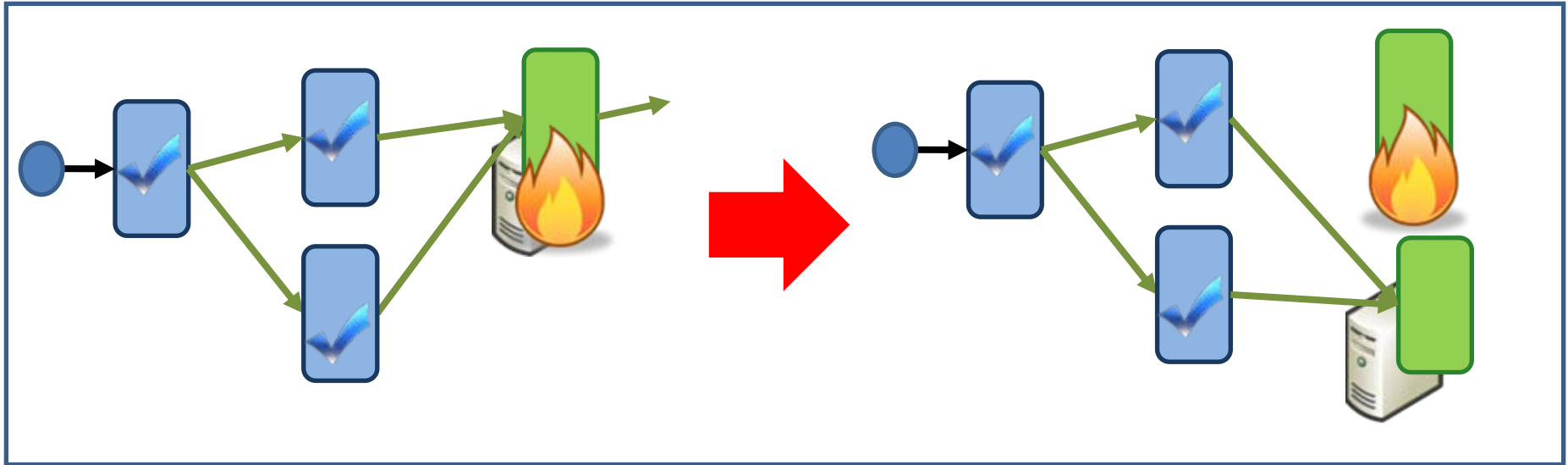
- 2D DAG
- multi-machine
- virtualized



# Architecture



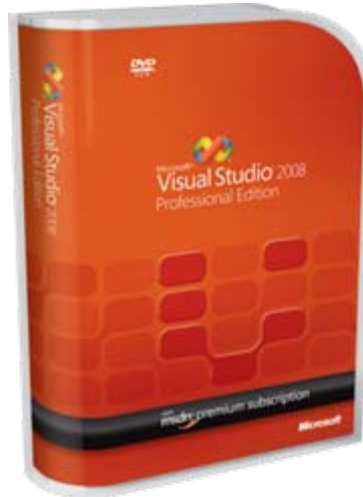
# Fault Tolerance



# Outline

- Introduction
- Dryad
- **DryadLINQ**
- **Applications**

# DryadLINQ



Dryad





# LINQ = C# + Queries

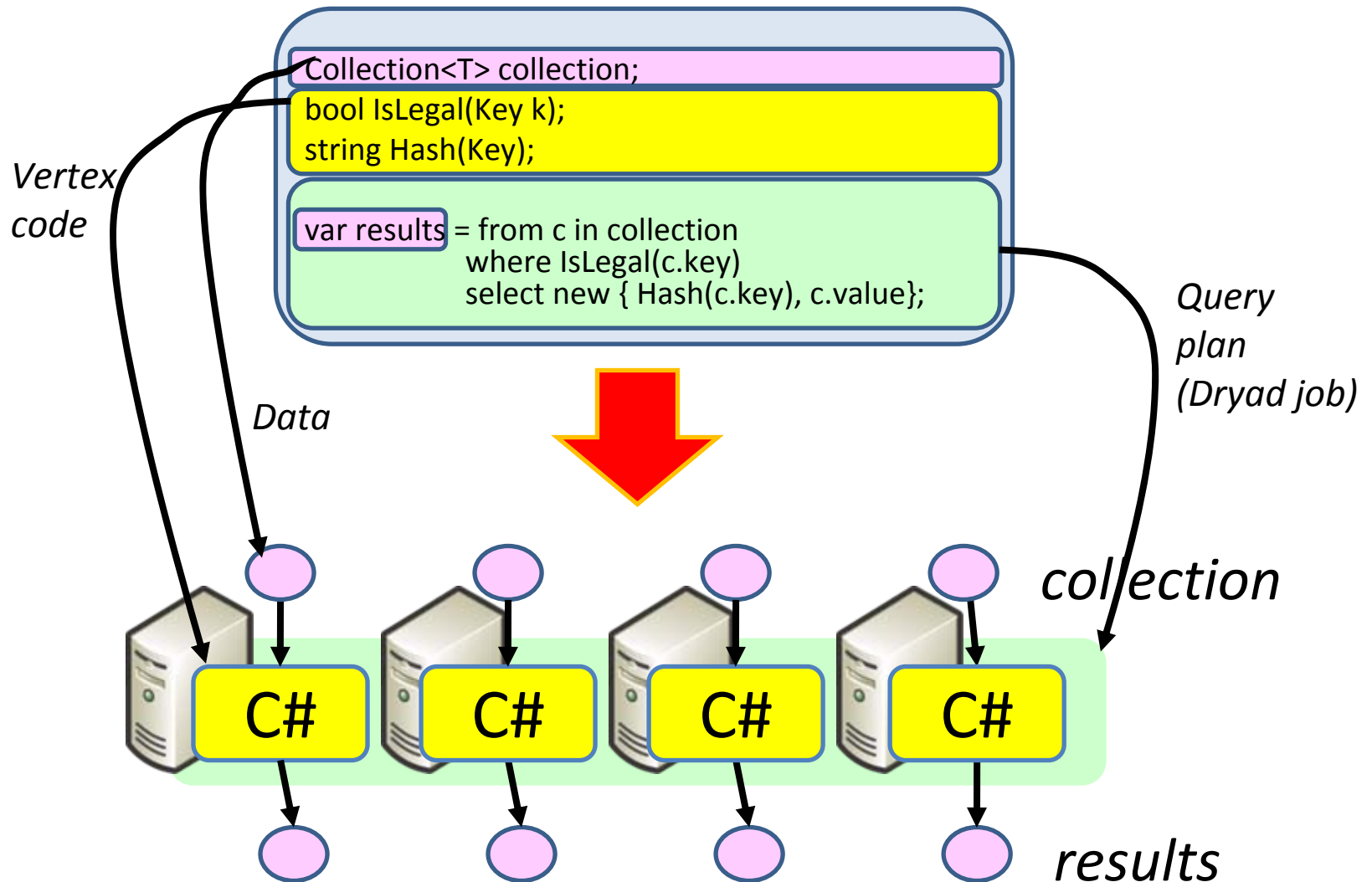
```
Collection<T> collection;
```

```
bool IsLegal(Key);
```

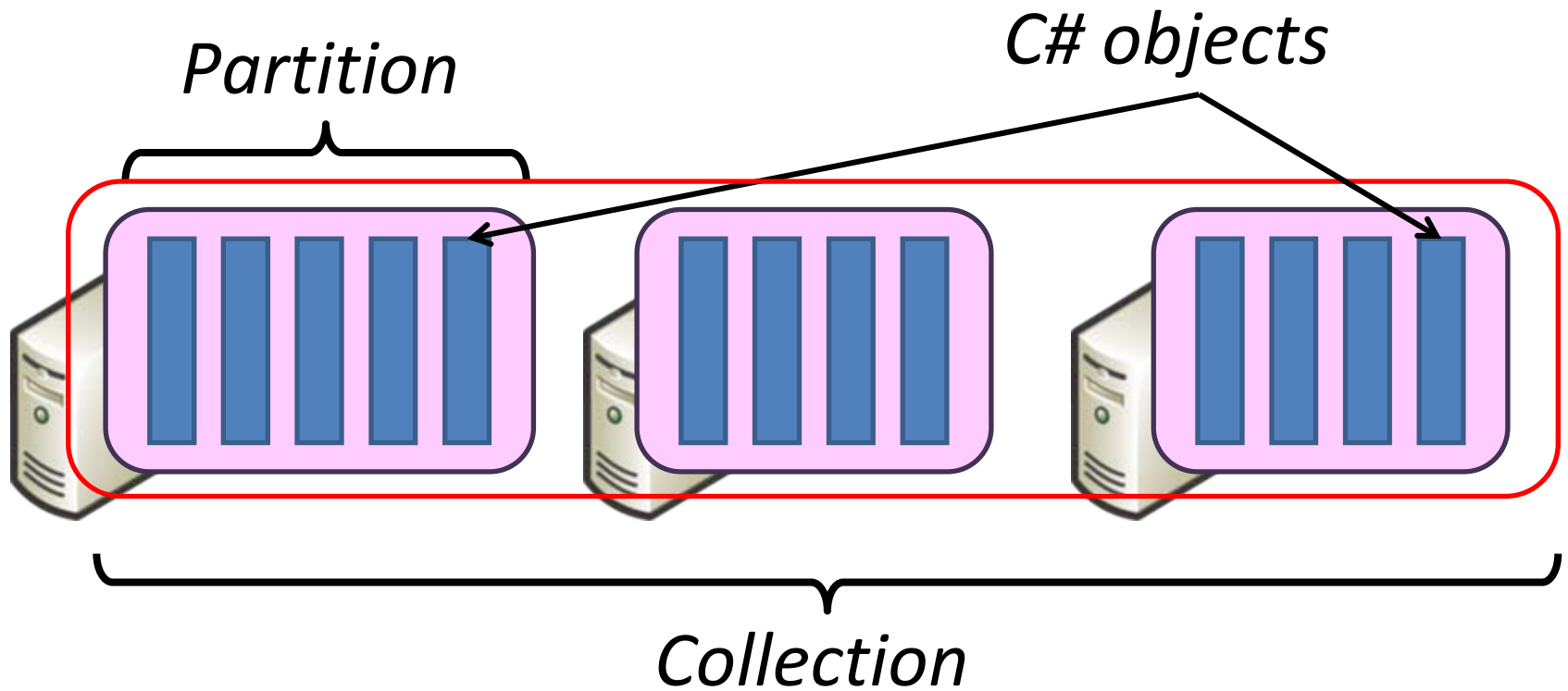
```
string Hash(Key);
```

```
var results = from c in collection  
              where IsLegal(c.key)  
              select new { Hash(c.key), c.value};
```

# DryadLINQ = LINQ + Dryad



# Data Model



# Language Summary


Where

Select

GroupBy

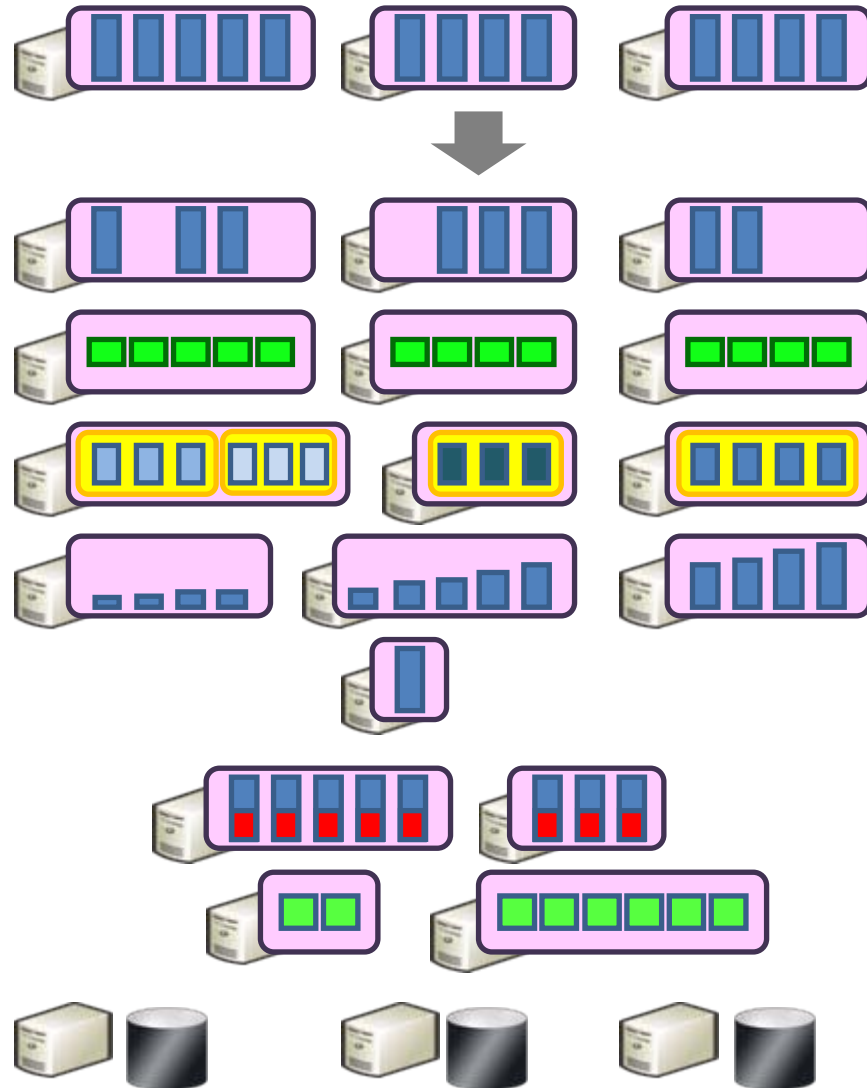
OrderBy

Aggregate

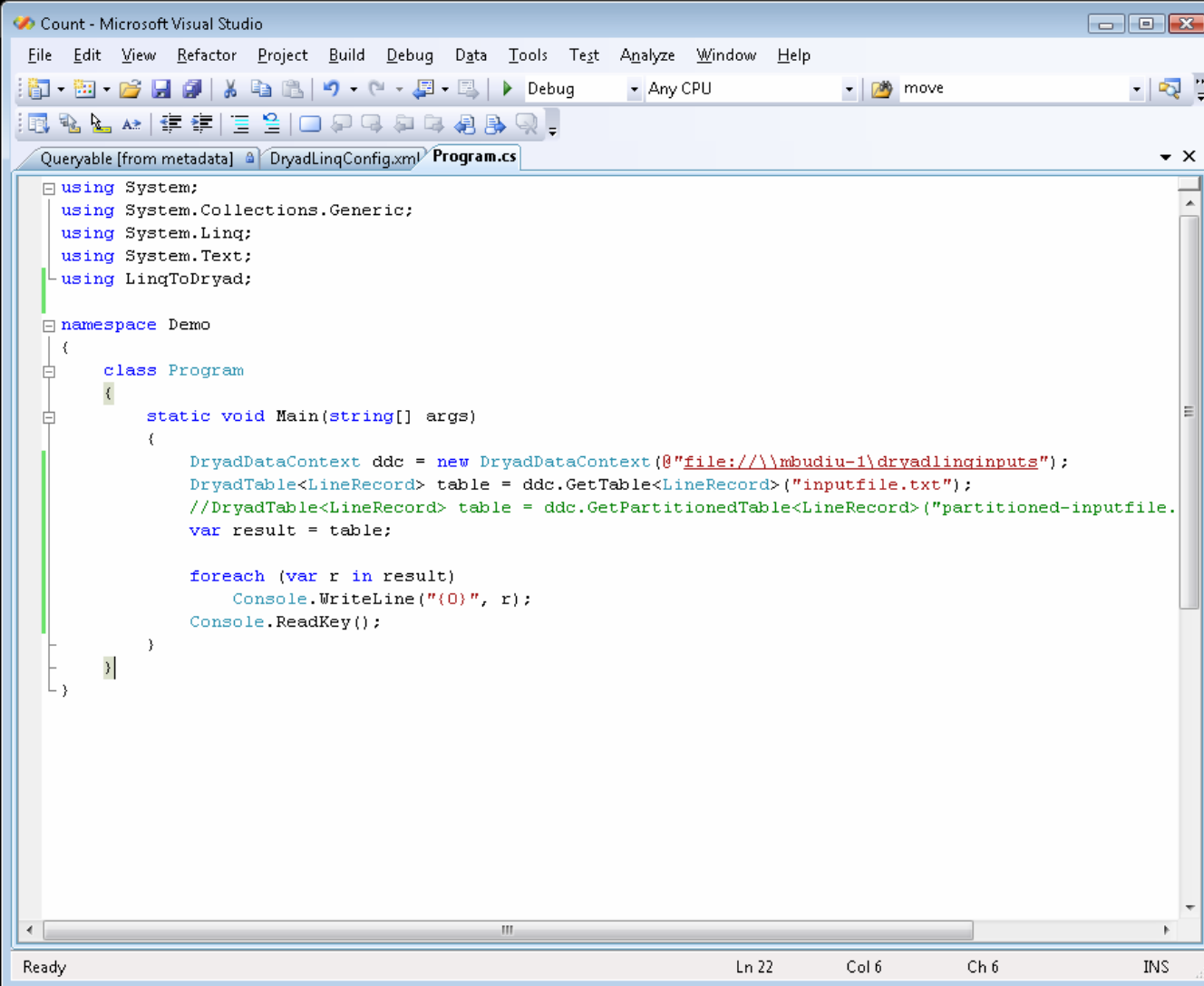
Join  →

Apply

Materialize



# Demo



```
using System;
using System.Collections.Generic;
using System.Linq;
using System.Text;
using LinqToDryad;

namespace Demo
{
    class Program
    {
        static void Main(string[] args)
        {
            DryadDataContext ddc = new DryadDataContext(@"file://\\mbudiu-1\dryadlinginputs");
            DryadTable<LineRecord> table = ddc.GetTable<LineRecord>("inputfile.txt");
            //DryadTable<LineRecord> table = ddc.GetPartitionedTable<LineRecord>("partitioned-inputfile.");
            var result = table;

            foreach (var r in result)
                Console.WriteLine("{0}", r);
            Console.ReadKey();
        }
    }
}
```

Ready Ln 22 Col 6 Ch 6 INS

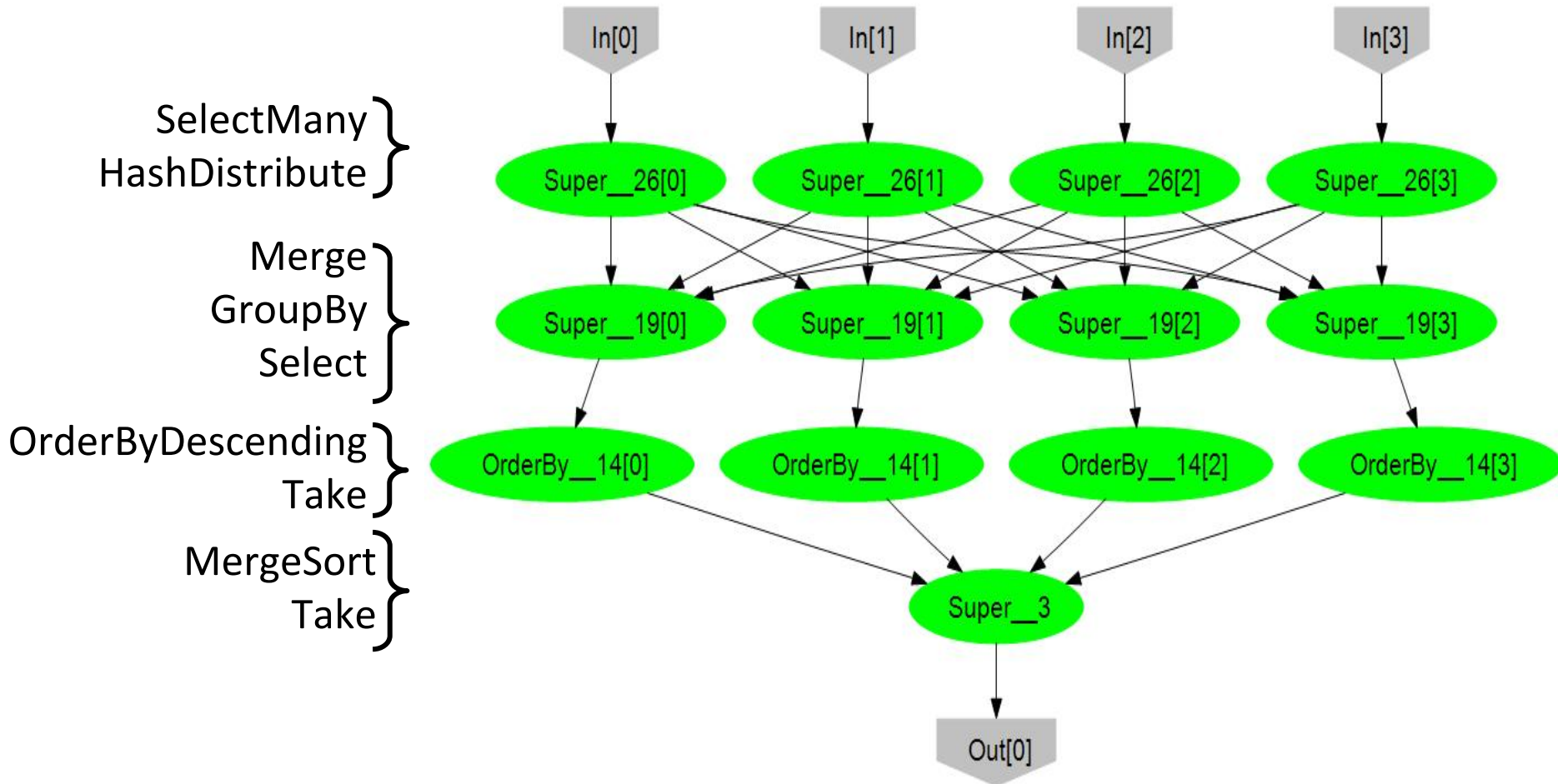
Done

# Example: Histogram

```
public static IQueryable<Pair> Histogram(
    IQueryable<LineRecord> input, int k)
{
    var words = input.SelectMany(x => x.line.Split(' '));
    var groups = words.GroupBy(x => x);
    var counts = groups.Select(x => new Pair(x.Key, x.Count()));
    var ordered = counts.OrderByDescending(x => x.count);
    var top = ordered.Take(k);
    return top;
}
```

"A line of words of wisdom"
["A", "line", "of", "words", "of", "wisdom"]
[["A"], ["line"], ["of", "of"], ["words"], ["wisdom"]]
[{"A", 1}, {"line", 1}, {"of", 2}, {"words", 1}, {"wisdom", 1}]
[{"of", 2}, {"A", 1}, {"line", 1}, {"words", 1}, {"wisdom", 1}]
[{"of", 2}, {"A", 1}, {"line", 1}]

# Histogram Plan

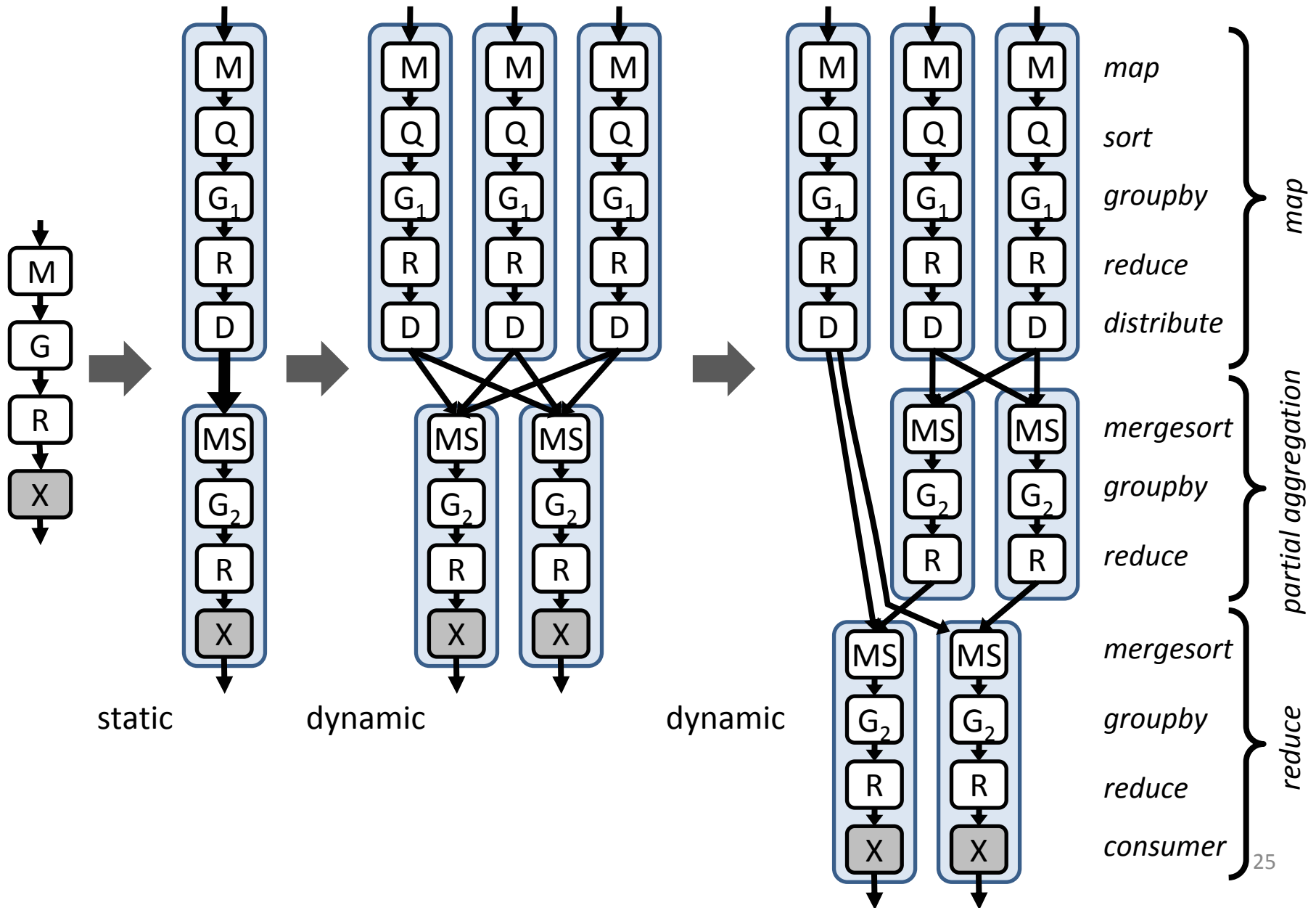


# Map-Reduce in DryadLINQ

```
public static IQueryable<S> MapReduce<T,M,K,S>(
    this IQueryable<T> input,
    Expression<Func<T, IEnumerable<M>>> mapper,
    Expression<Func<M,K>> keySelector,
    Expression<Func<IGrouping<K,M>,S>> reducer)
{
    var map = input.SelectMany(mapper);
    var group = map.GroupBy(keySelector);
    var result = group.Select(reducer);
    return result;
}
```



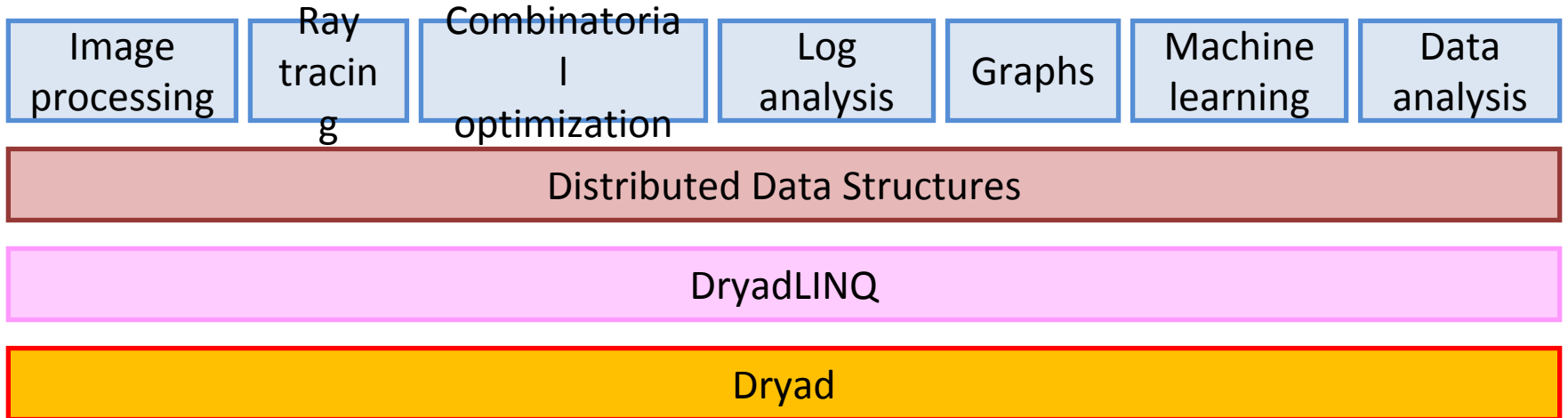
# Map-Reduce Plan



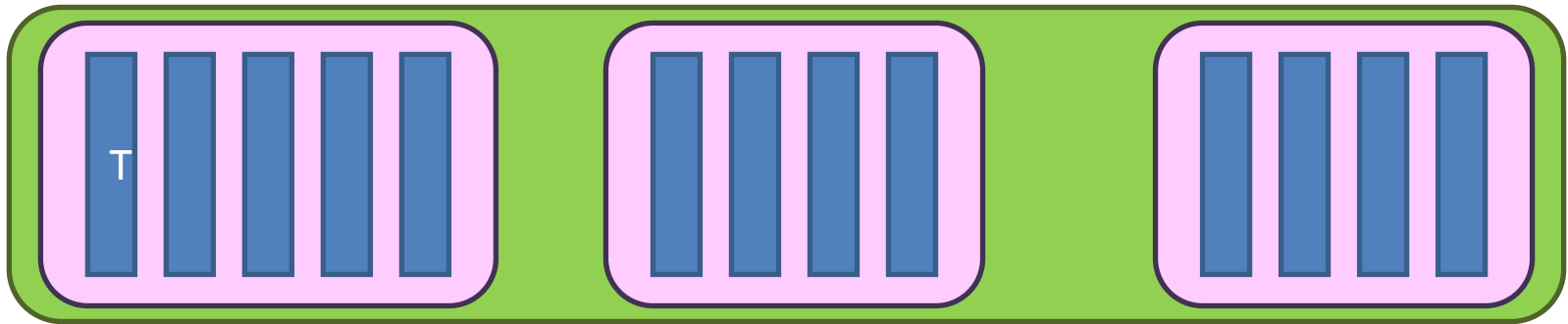
# Outline

- Introduction
- Dryad
- DryadLINQ
- **Applications**

# Applications

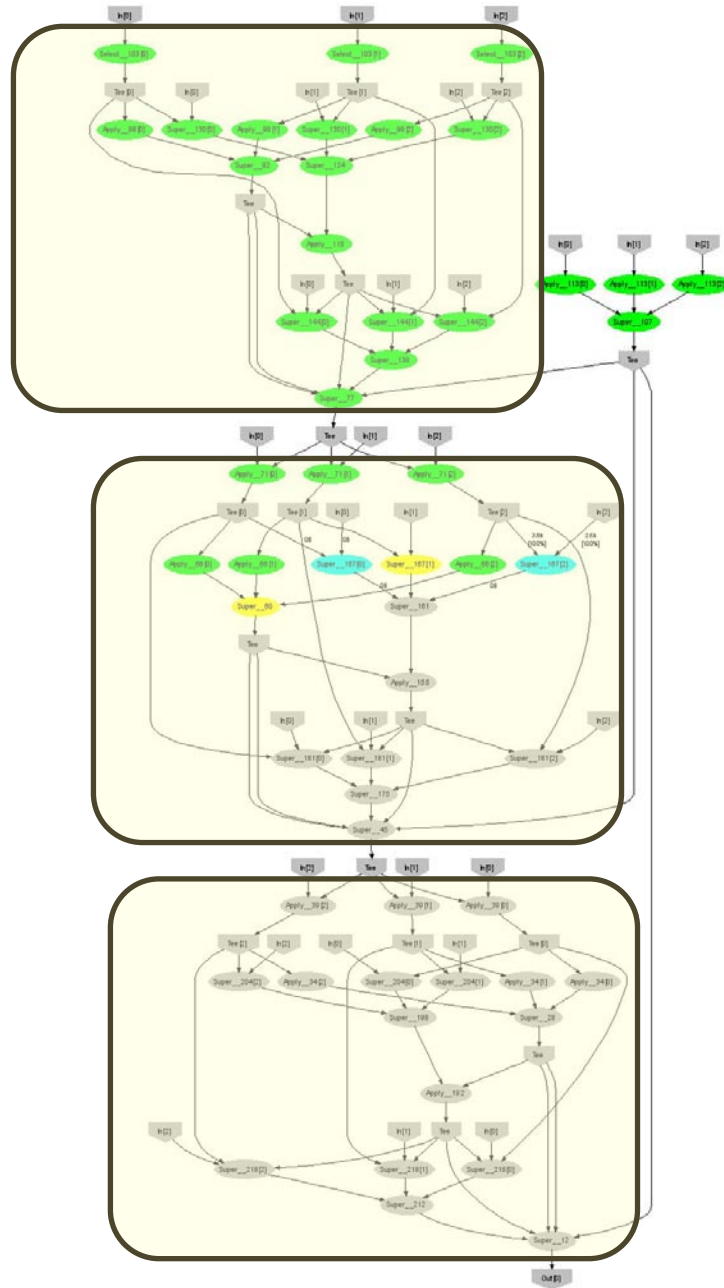


# E.g: Linear Algebra



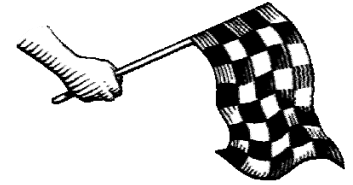
$$\left\{ \begin{array}{c} \text{blue bar} \\ \text{T} \end{array}, \begin{array}{c} \text{yellow bar} \\ \text{U} \end{array}, \begin{array}{c} \text{red bar} \\ \text{V} \end{array} \right\} = \left\{ \mathcal{R}, \mathcal{R}^m, \mathcal{R}^{m \times n} \right\}$$

# Expectation Maximization (Gaussians)



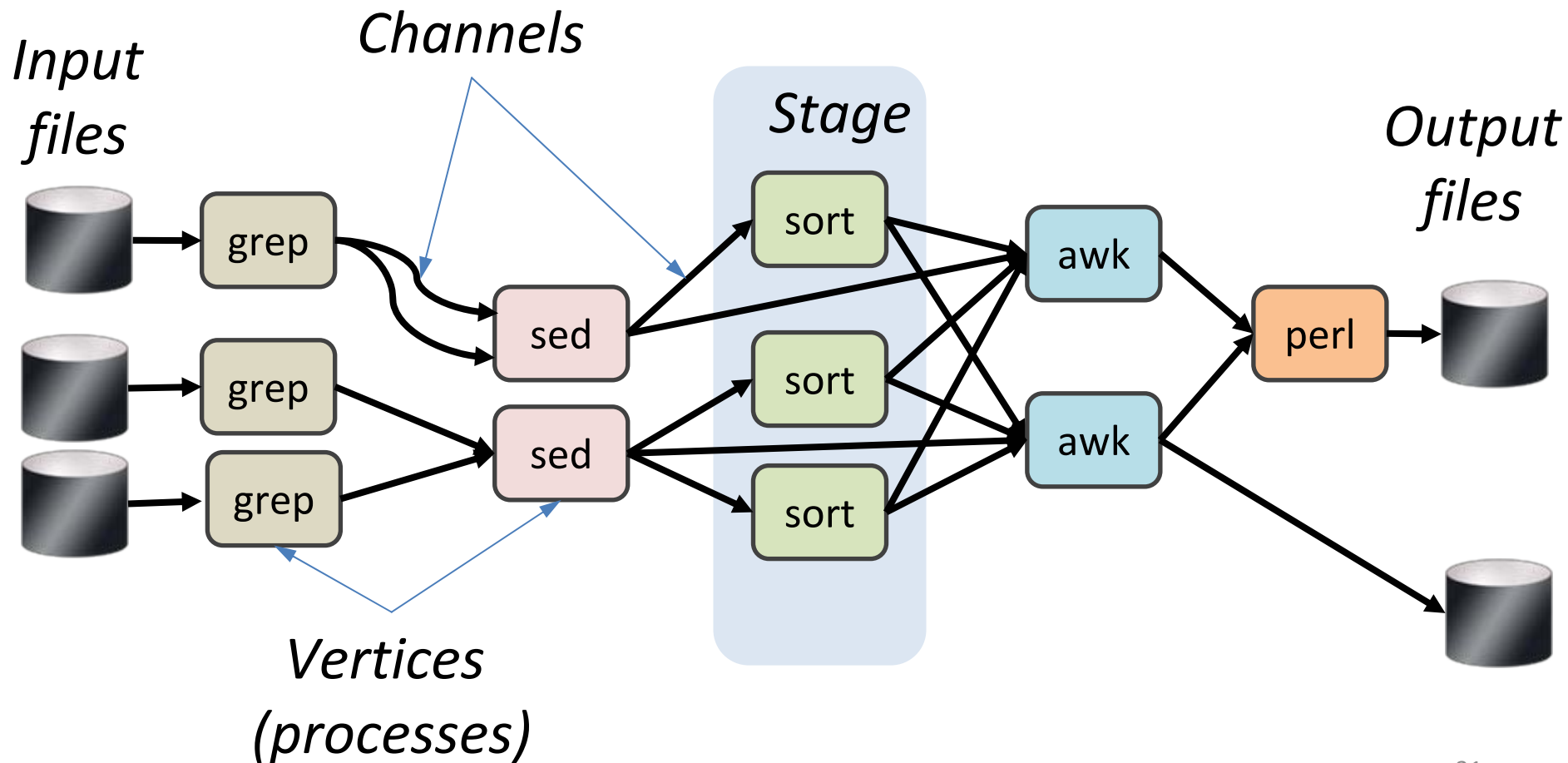
- 160 lines
- 3 iterations shown

# Conclusions



- Dryad = distributed execution environment
- Supports rich software ecosystem
  
- DryadLINQ = Compiles LINQ to Dryad
- C# objects and declarative programming
- .Net and Visual Studio  
for distributed programming

# Dryad Job Structure



# Linear Regression

- Data

$$x_t \in \mathcal{R}^n, y_t \in \mathcal{R}^m \quad t \in \{1, \dots, n\}$$

- Find

$$A \in \mathcal{R}^{n \times m}$$

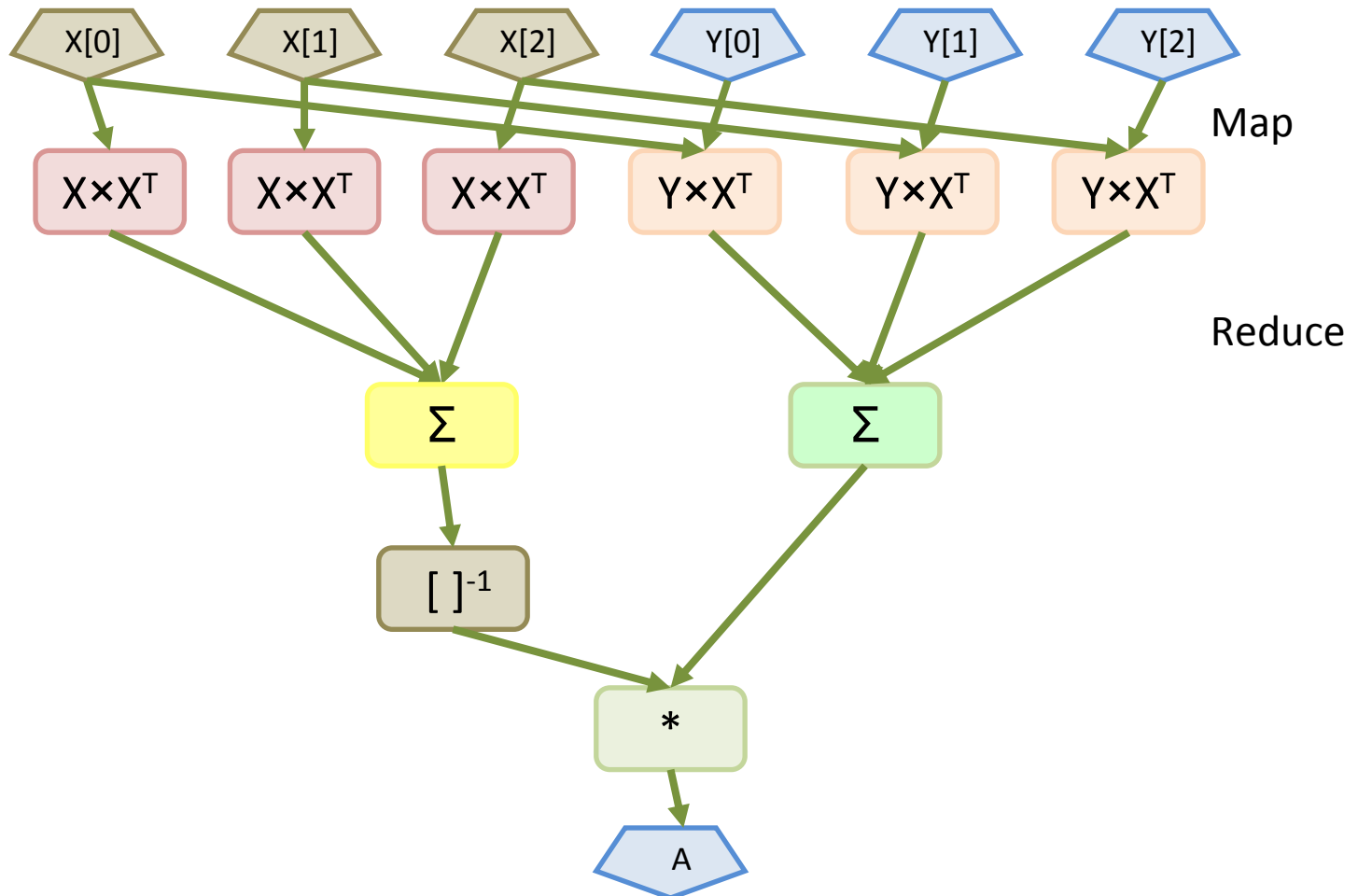
- S.t.

$$Ax_t \approx y_t$$



# Analytic Solution

$$A = \left( \sum_t y_t \times x_t^T \right) \left( \sum_t x_t \times x_t^T \right)^{-1}$$



# Linear Regression Code

$$A = \left( \sum_t y_t \times x_t^T \right) \left( \sum_t x_t \times x_t^T \right)^{-1}$$

Vectors  $x = \text{input}(0)$ ,  $y = \text{input}(1)$ ;

Matrices  $xx = x.\text{Map}(x, (a,b) \Rightarrow a.\text{OuterProd}(b))$ ;

OneMatrix  $xxs = xx.\text{Sum}()$ ;

Matrices  $yx = y.\text{Map}(x, (a,b) \Rightarrow a.\text{OuterProd}(b))$ ;

OneMatrix  $yxs = yx.\text{Sum}()$ ;

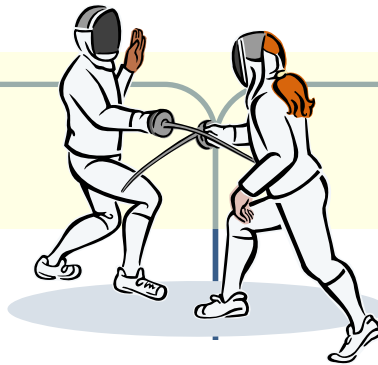
OneMatrix  $xxinv = xxs.\text{Map}(a \Rightarrow a.\text{Inverse}())$ ;

OneMatrix  $A = yxs.\text{Map}(xxinv, (a, b) \Rightarrow a.\text{Mult}(b))$ ;

# Dryad = Execution Layer



# Dryad



# Map-Reduce

- Many similarities

- Execution layer
- Job = arbitrary DAG
- Plug-in policies
- Program=graph gen.
- Complex (↑features)
- New (< 2 years)
- Still growing
- Internal

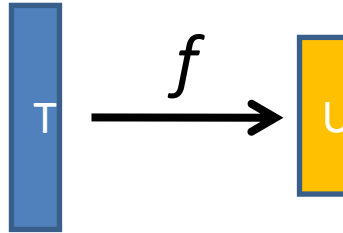
- Exe + app. model
- Map+sort+reduce
- Few policies
- Program=map+reduce
- Simple
- Mature (> 4 years)
- Widely deployed
- Hadoop

# PLINQ

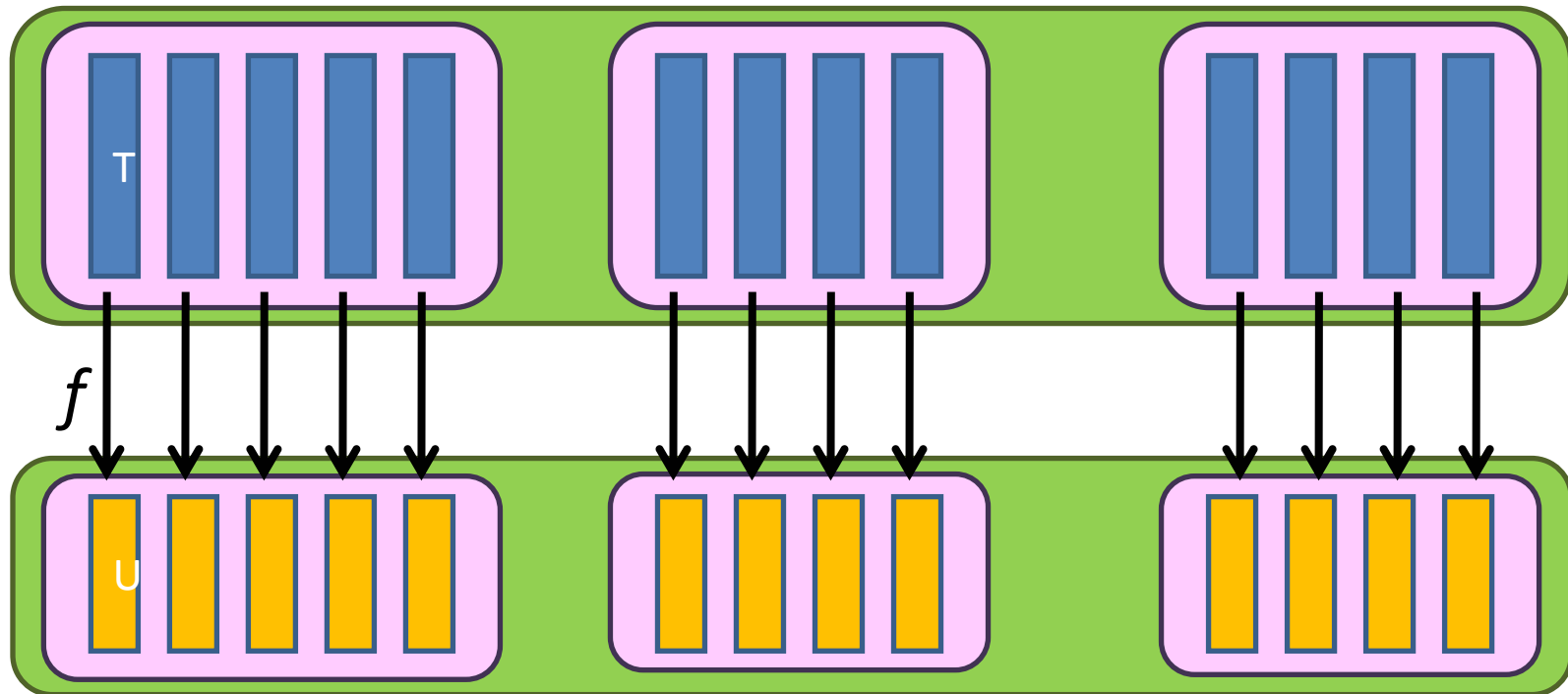
```
public static IEnumerable<TSource>  
    DryadSort<TSource, TKey>(IEnumerable<TSource> source,  
                             Func<TSource, TKey> keySelector,  
                             IComparer<TKey> comparer,  
                             bool isDescending)  
{  
    return source.AsParallel().OrderBy(keySelector, comparer);  
}
```

# Operations on Large Vectors:

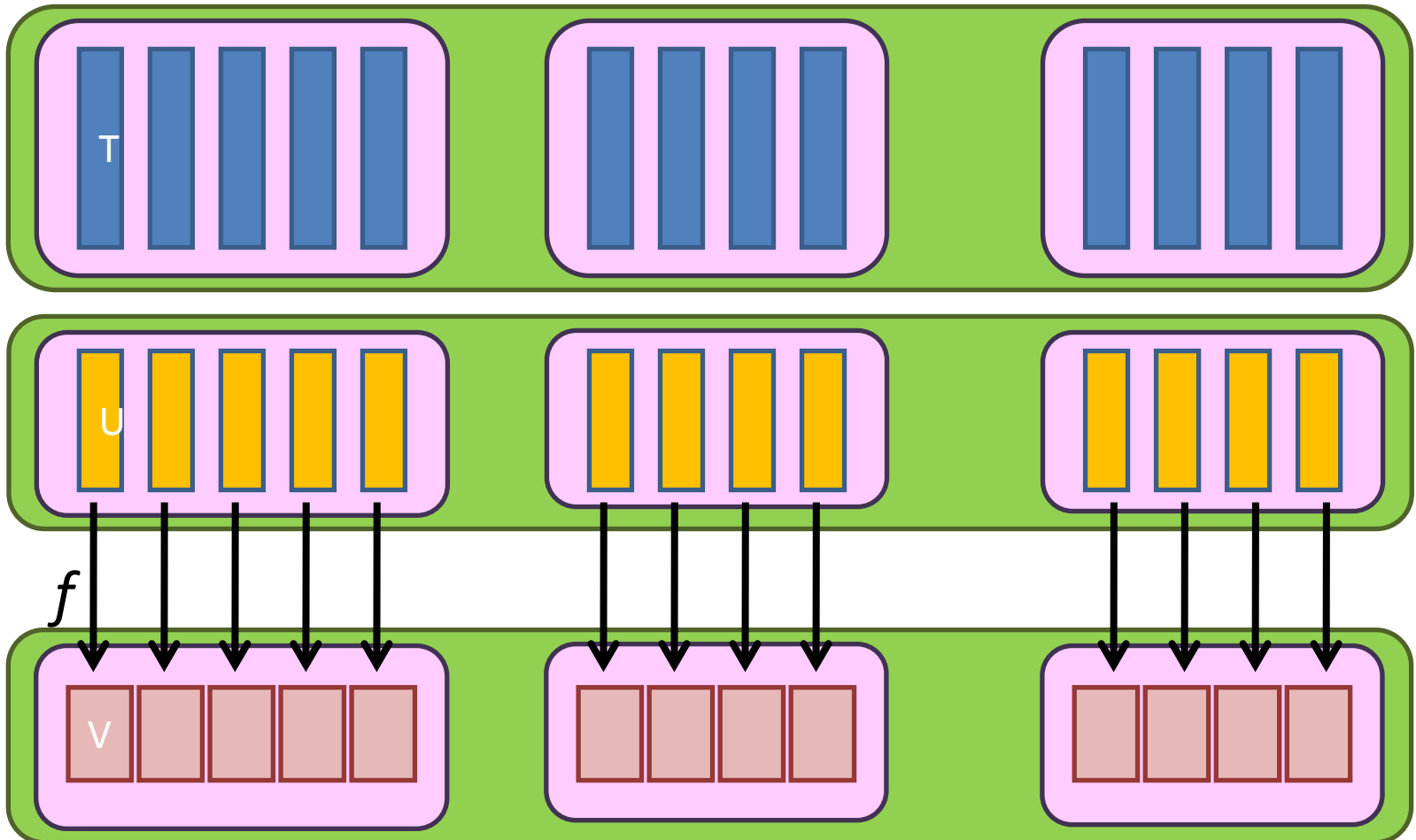
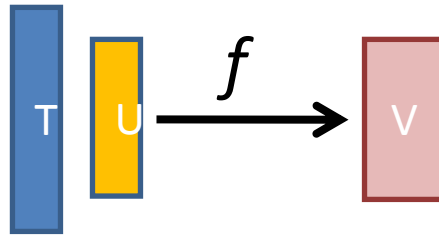
## Map 1



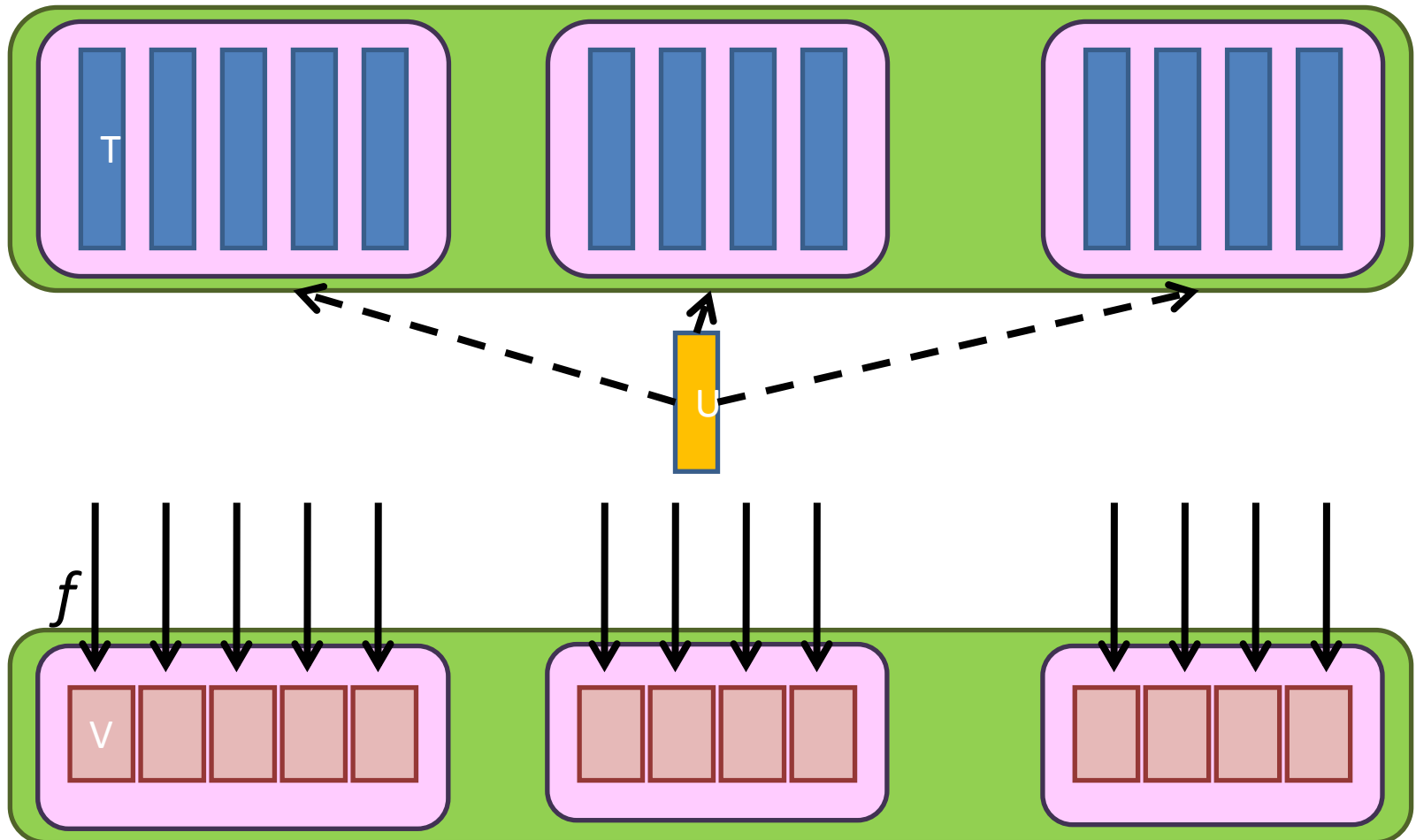
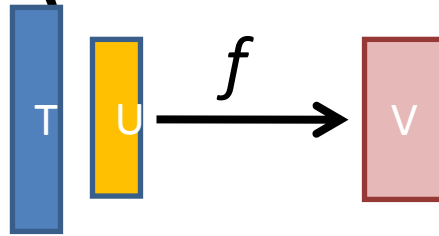
$f$  preserves partitioning



# Map 2 (Pairwise)

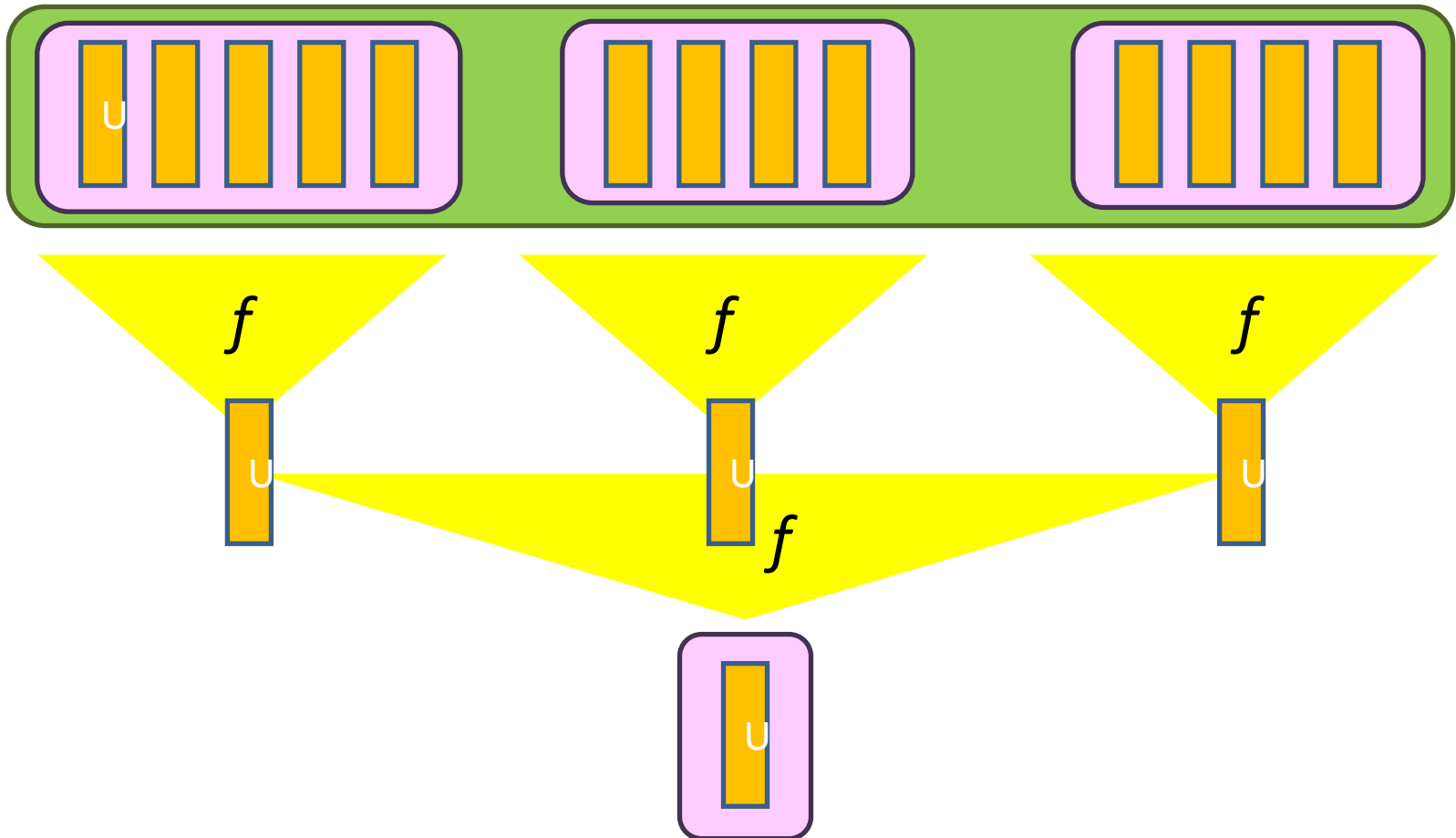
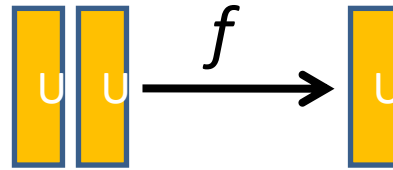


# Map 3 (Vector-Scalar)





# Reduce (Fold)



# Software Stack

