

IBM Almaden Research Center



# Jaql

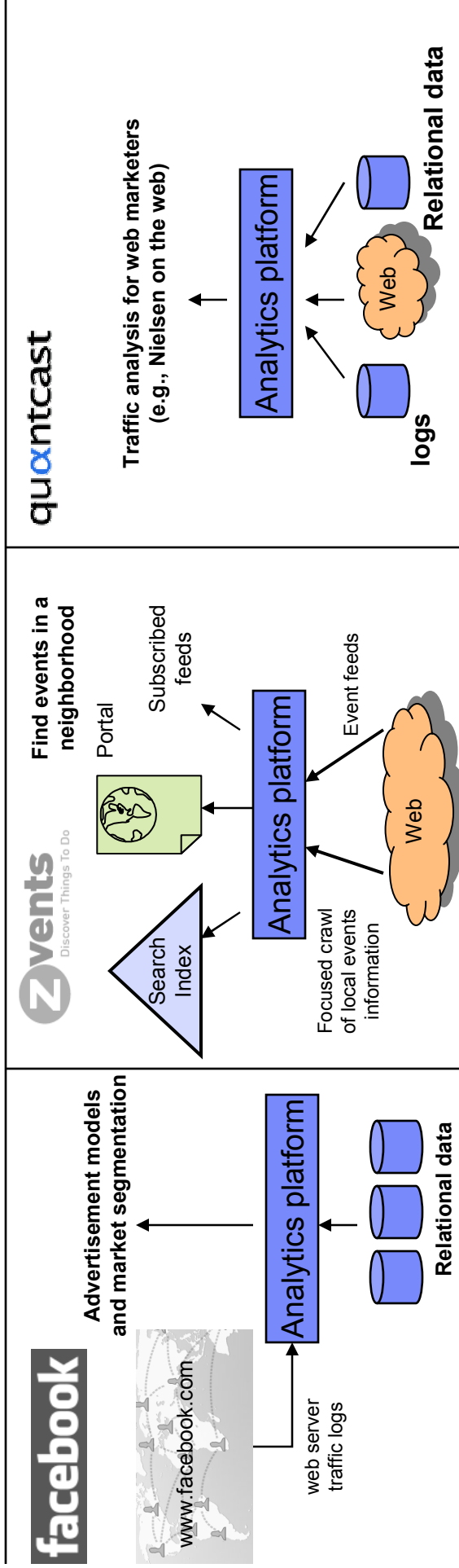
Running Pipes in the Clouds

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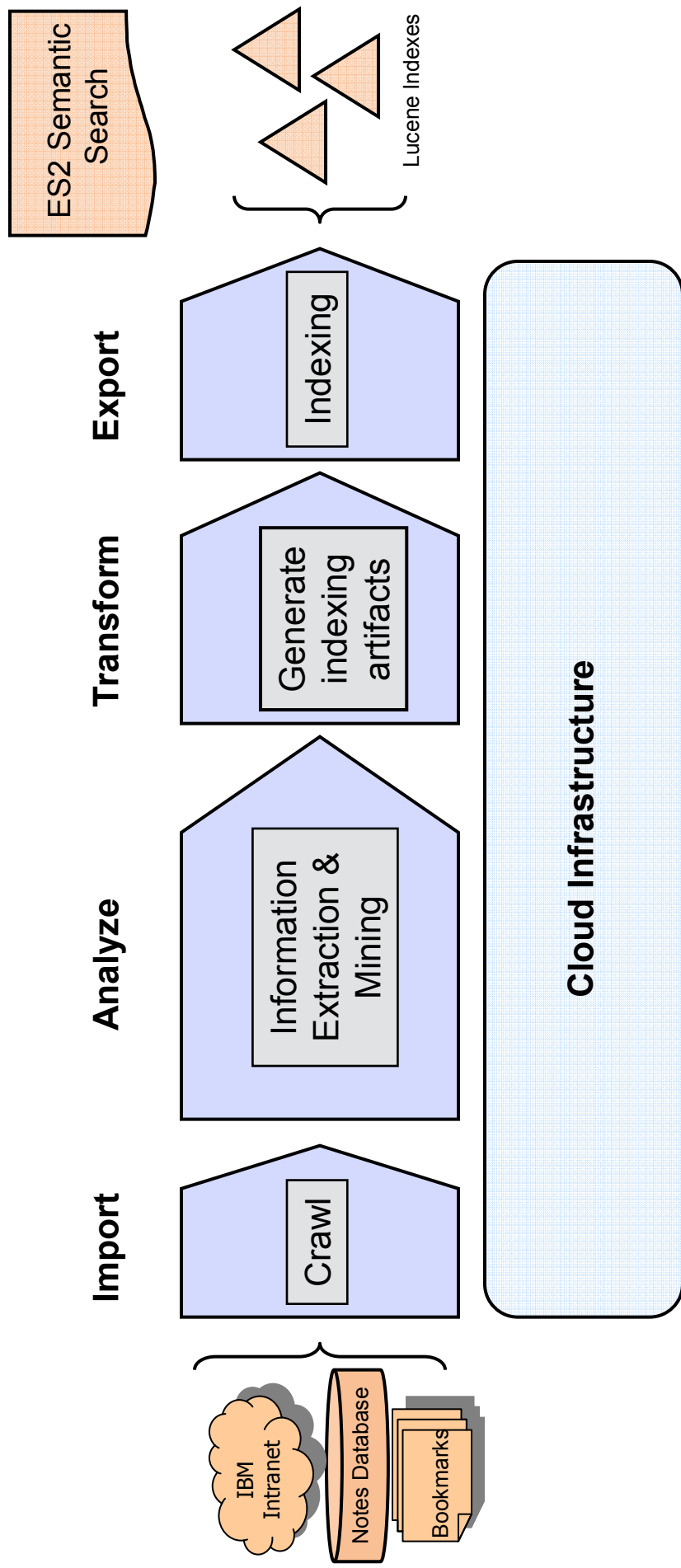
<http://code.google.com/p/jaql/>

# Motivating Scenarios

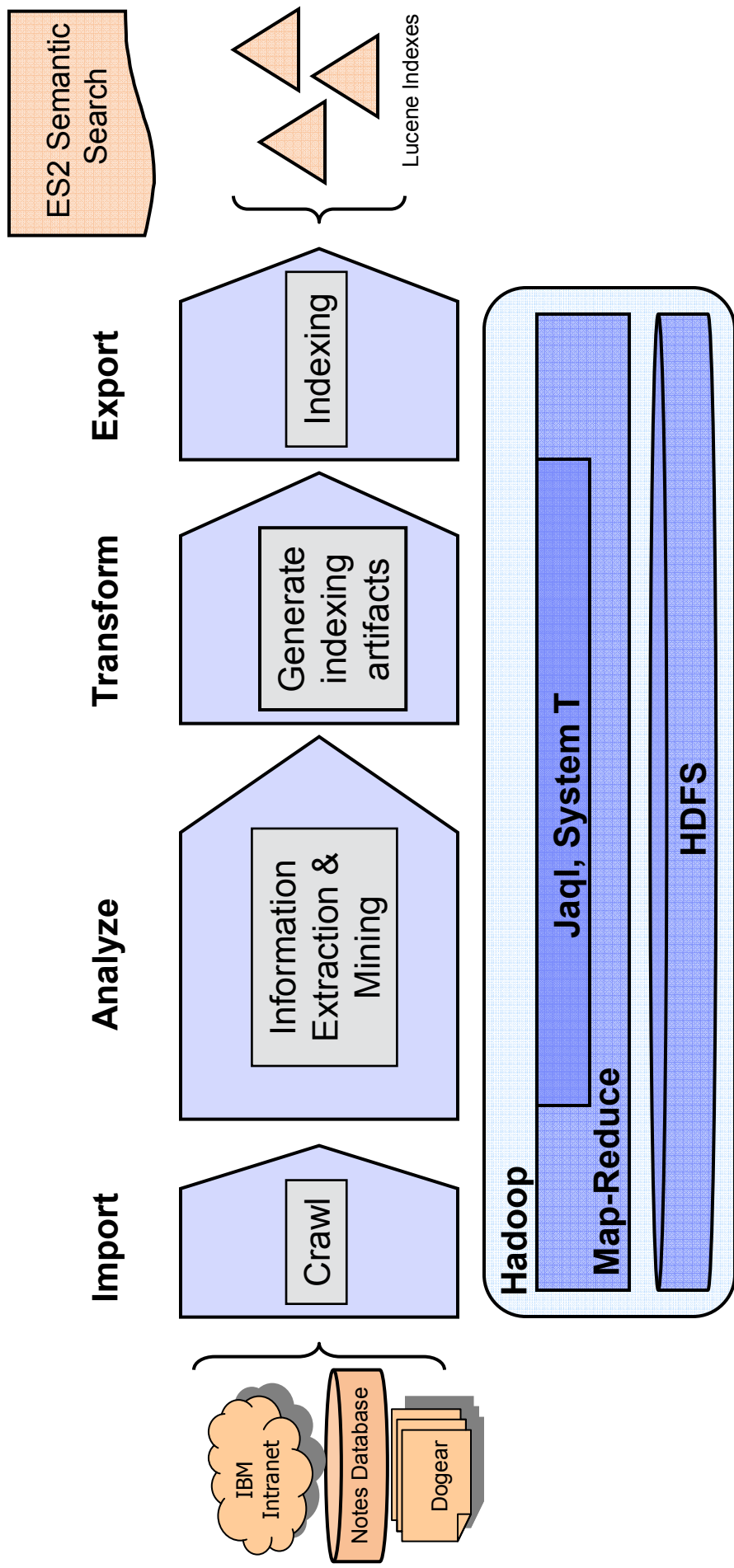
- Driving forces
  - Data characteristics
    - Unstructured/Semi-structured (e.g., call-center records, pathology reports, ..)
    - Dynamic & Continuously Evolving (e.g., click streams, application logs, system logs)
    - Massive Scale (several terabytes to petabytes ..)
  - Nature of analytics
    - Compute intensive
    - Evolving
- Demands a **flexible** platform to dynamically accommodate new analytic flows



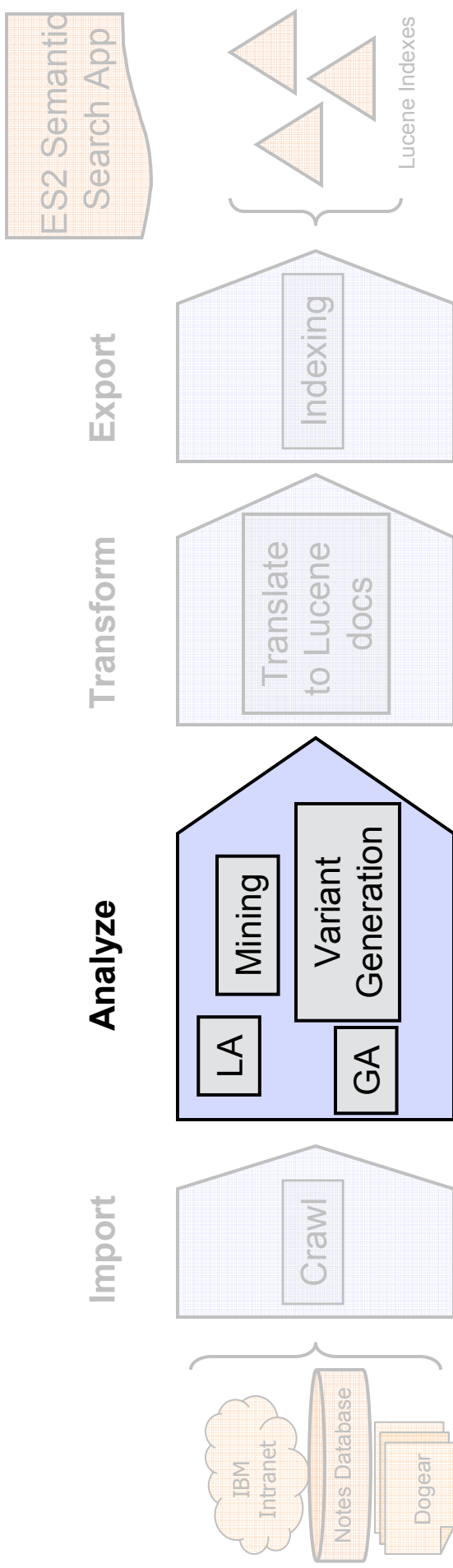
# ES2: Intranet Data → Semantic Search



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# ES2's Analysis Stage



**Local Analysis:** **System T** extracts structured information, e.g., candidate home pages w/ *person* name

**Global Analysis:** **Jaql** finds the best pages within groups of pages, e.g., *personal homepage*

**Mining:** Use **map-reduce** for mining

- Acronym extraction
- Geo-classification
- Learning regular expressions

## Goals for Jaql

- Semi-structured data manipulation
  - Use JSON as a data model
  - Data is converted to/from JSON view
  - Most data has a natural JSON representation
    - Relational tables, CSV, XML, ...
- Easily extended
  - E.g., Java, Python, JavaScript, ...
- Exploit massive parallelism using Hadoop
  - Queries compiled to map-reduce jobs

## JSON Examples

[ ] == array, { } == record or object, xxx: == field name

```
[
  { from:17, to: 19, msg:" ... ",
    phones: ["+1-415-555-1212", "+1-408-555-1234"], names:["Jill", "Jenny"] },
  { from: 17, to: 12, msg:" ... ", names:["Jack", "Jill"]},
  { from: 17, to: 12, msg:" ... ", phones: ["+1-415-555-2345"] }
]
```

Message Log

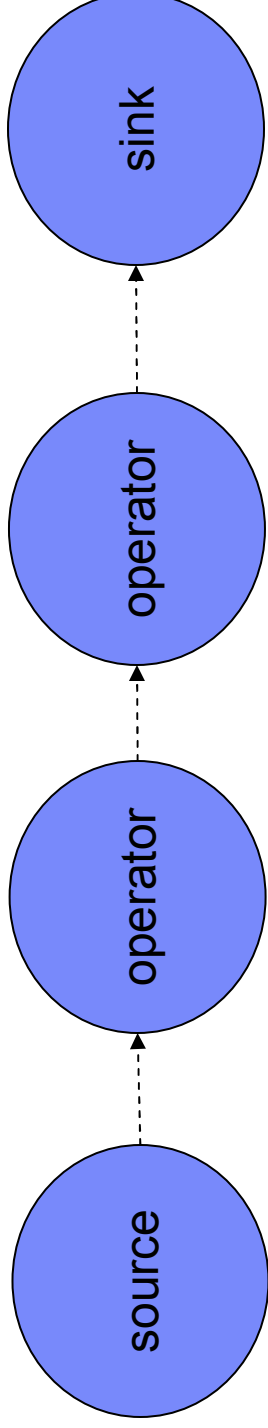
```
[
  { id: 12, name: "Joe Smith", bday: date("1971-03-07"), zip: 94114 },
  { id: 17, name: "Ann Jones", bday: date("1973-02-04"), zip: 94110 },
  { id: 19, name: "Alicia Fox", bday: date("1975-04-20"), zip: 94114 }
]
```

User Info

```
[
  { city: "San Francisco", zips: [94110,94112,94114], areacodes: [415] },
  { city: "San Jose", zips: [95120,95123], areacodes: [408] }
]
```

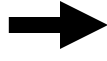
City Info

## Writing a pipeline in Jaql



### Find users in zip 94114

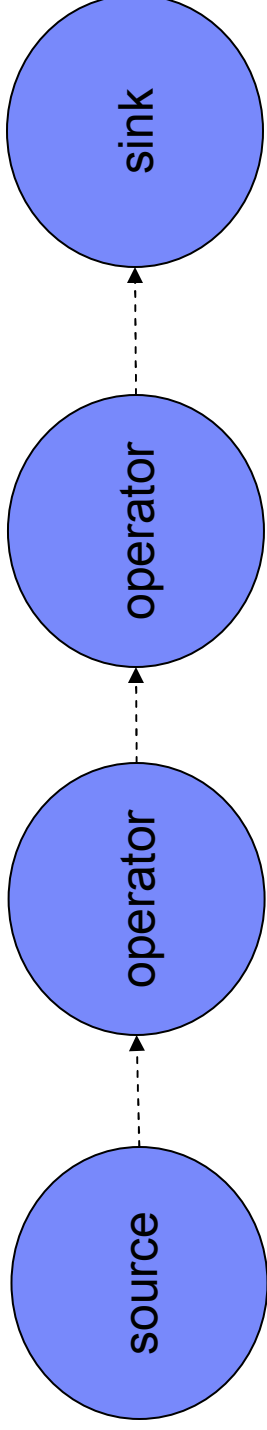
```
[
  { id: 12, name: "Joe Smith", bday: date("1971-03-07"), zip: 94114 },
  { id: 17, name: "Ann Jones", bday: date("1973-02-04"), zip: 94110 },
  { id: 19, name: "Alicia Fox", bday: date("1975-04-20"), zip: 94114 }
]
```



```
[
  { id: 12, name: "Joe Smith" },
  { id: 19, name: "Alicia Fox" }
]
```



## Writing a pipeline in Jaql: read



### Find users in zip 94114

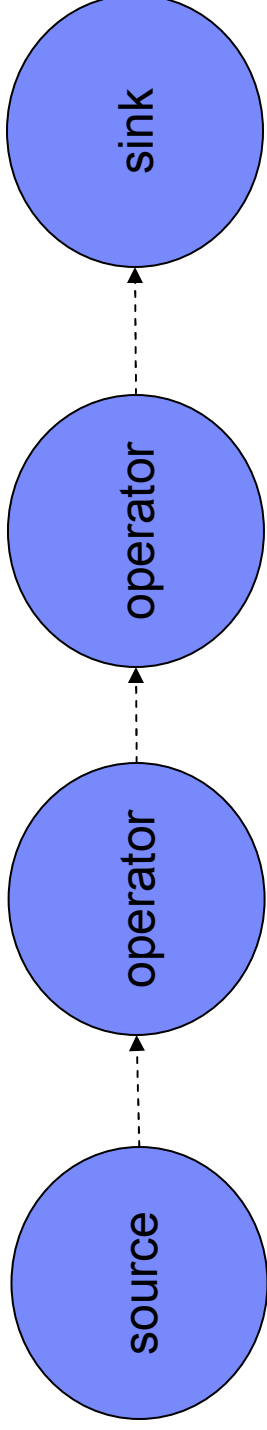
#### Query

```
read("hdfs:users");
```

#### Data

```
[
  { id: 12, name: "Joe Smith",
    bday: date("1971-03-07"), zip: 94114 },
  { id: 17, name: "Ann Jones",
    bday: date("1973-02-04"), zip: 94110 },
  { id: 19, name: "Alicia Fox",
    bday: date("1975-04-20"), zip: 94114 }
]
```

## Writing a pipeline in Jaql: filter



### Find users in zip 94114

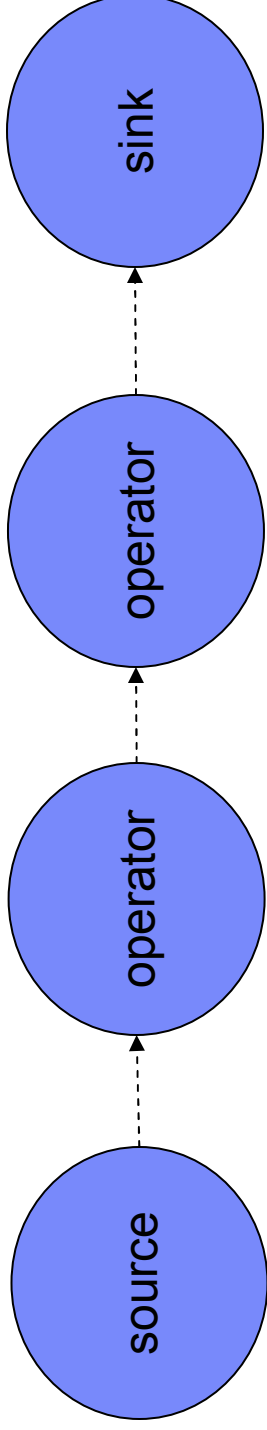
#### Query

```
read("hdfs:users")
-> filter $.zip == 94114;
```

#### Data

```
[
  { id: 12, name: "Joe Smith",
    bday: date("1971-03-07"), zip: 94114 },
  { id: 19, name: "Alicia Fox",
    bday: date("1975-04-20"), zip: 94114 }
]
```

## Writing a pipeline in Jaql: transform



### Find users in zip 94114

#### Query

```

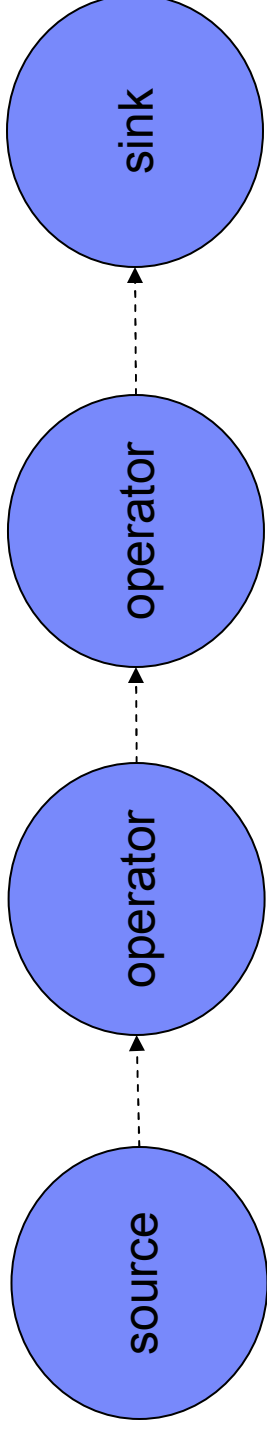
read("hdfs:users")
-> filter $.zip == 94114
-> transform { $.id, $.name };
  
```

#### Data

```

[
  { id: 12, name: "Joe Smith" },
  { id: 19, name: "Alicia Fox" }
]
  
```

## Writing a pipeline in Jaql: write



### Find users in zip 94114

#### Query

```

read("hdfs:users")
-> filter $.zip == 94114
-> transform { $.id, $.name }
-> write("hdfs:inzip");
  
```

#### Data

```

[
  { id: 12, name: "Joe Smith" },
  { id: 19, name: "Alicia Fox" }
]
  
```

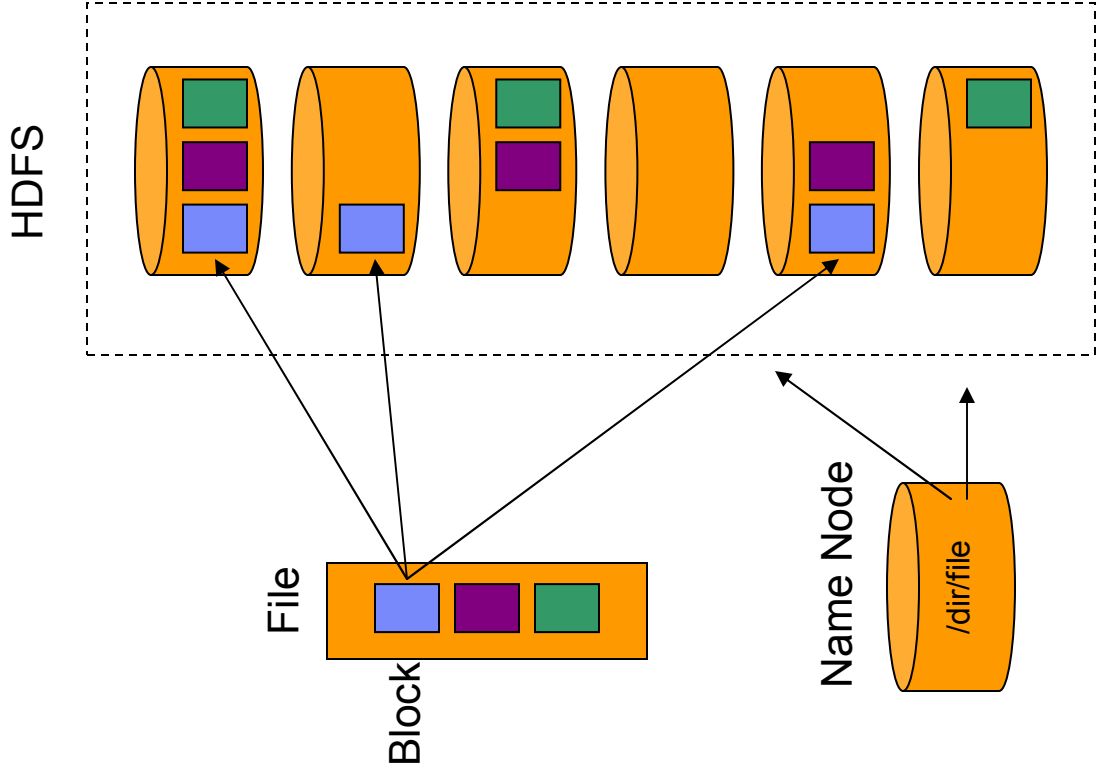
# JAQL I/O



- I/O layer abstracts details of data location + format
  - Examples of data stores:
    - HDFS, HBase, Amazon's S3, local FS, HTTP request, JDBC call
  - Examples of data formats:
    - JSON text, CSV, XML (default is JSON binary)
- JSON used as a data model

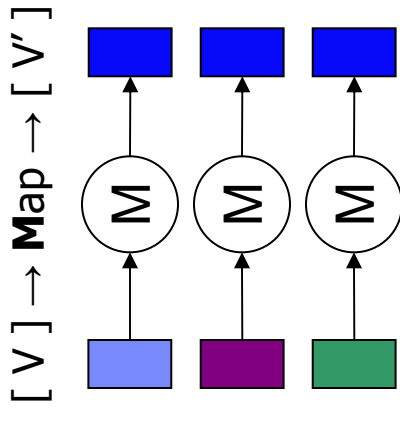
## Storing Files in HDFS

- File is divided into blocks
  - Default block is 64MB
- Blocks are replicated on HDFS nodes
  - Default replication is three times
- The name node tracks file names and block locations



# Map Jobs in Hadoop

- **JobConf**
  - Specifies job configuration
- **InputFormat**
  - Abstract data source
  - Creates Splits and RecordReader
- **Split**
  - Abstract notion of a data chunk
  - Each split handled by a map task
  - Contains location hints
- **RecordReader**
  - Produces stream of values from a Split
- **Map Task**
  - Processes values in one Split
  - Calls user map code
  - Reevaluated on error
- **OutputFormat**
  - Abstract data sink



- **Infrastructure handles**
  - Parallelism
  - Load balancing
  - Fault-tolerance

## Jaql using Map

```
read("hdfs:users")
-> filter $.zip == 94114
-> transform { $.id, $.name }
-> write("hdfs:inzip");
```

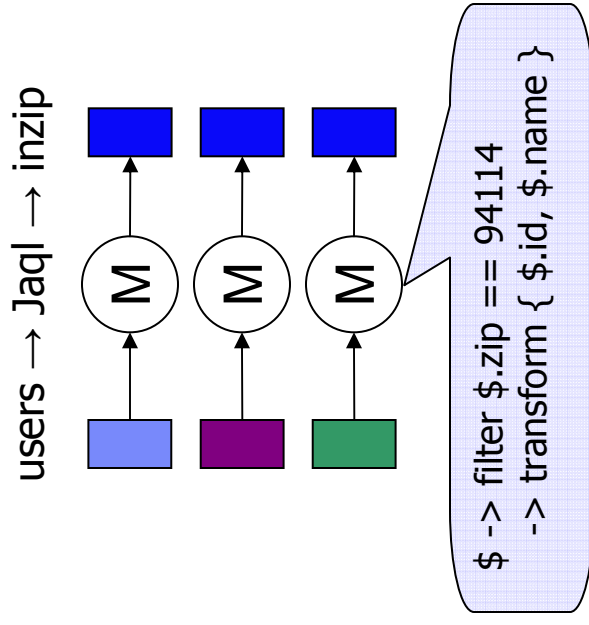


Rewrite Engine



## Equivalent map-reduce job in Jaql

```
mapReduce({
  input : {uri: "hdfs:users" },
  map   : fn($) ( $ -> filter $.zip == 94114
                 -> transform { $.id, $.name } ),
  output : { uri: "hdfs:inzip" } })
```





## User-defined Functions

Identify phone numbers in each message

```
read("hdfs:rawMsgs")
-> transform {$.*, phones: findPhones($.msg)};
```

Express user code with:

- Java
- Inline scripting
- ...

Input data

```
[
  { from:17, to: 19, msg:"..." }
  { from: 17, to: 12, msg:"..." },
  { from: 17, to: 19, msg:"..." }
]
```



Output data

```
[
  { from:17, to: 19, msg:"..." },
    phones: ["+1-415-555-1212",
            "+1-408-555-1234"] },
  { from: 17, to: 12, msg:"..." },
  { from: 17, to: 19, msg:"..." },
    phones: ["+1-415-555-2345"] }
]
```

## Inline Scripting

```
script python <<END

import re
import phoneUtil

phoneRE = re.compile('[+x]?[0-9\-(\)]+')

def findPhones(text):
    for p in phoneRE.findall(text):
        p = phoneUtil.standardize(p)
        if p != None:
            yield p

END;

read("hdfs:rawMsgs")
-> transform {$.*, phones: findPhones($.msg)};
```

- Easily integrate with other programming languages
  - Python first
  - then JavaScript, Ruby, Perl, Unix shell, etc

## Grouping

Find the number of times the each sender mentioned an area code

```
read("\hdfs:rawMsgs")
-> expand for( $p in findPhones($.msg) ) [{ $.from, area: area($p) }]
-> group by $g = ({$$.from, $.area})
into { $g.*, n: count($) };
```

Input data

```
[
  { from:17, to: 19, msg:" ..." }
  { from: 17, to: 12, msg:" ..." },
  { from: 17, to: 19, msg:" ..." }
]
```

After expand

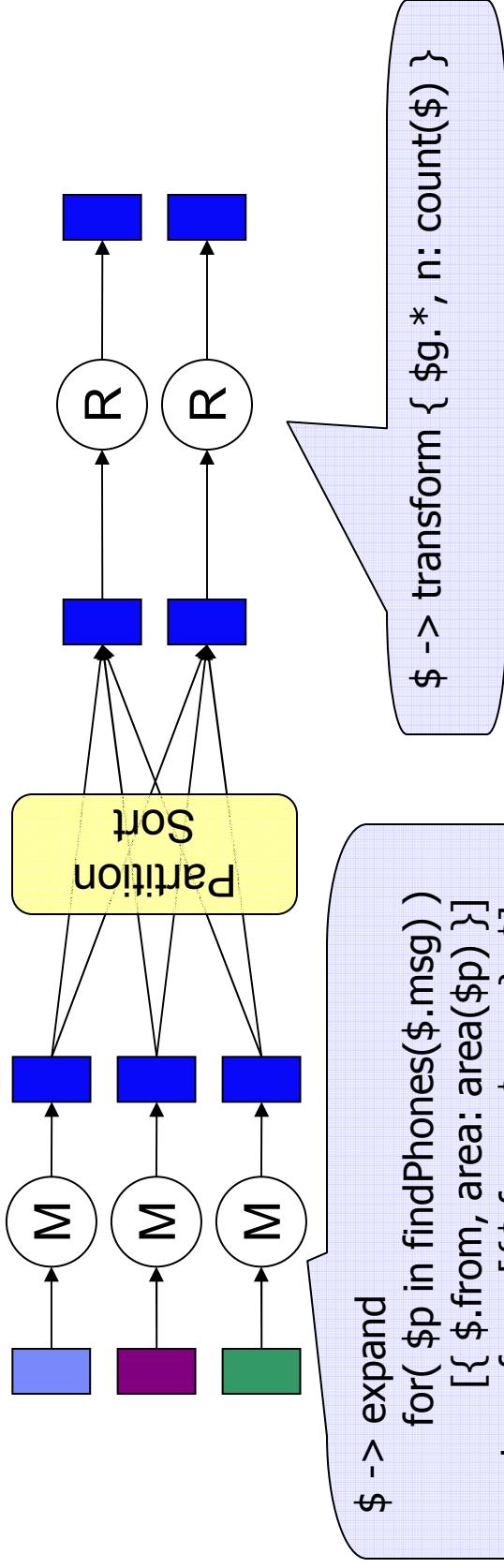
```
[
  { from: 17, area: 415 },
  { from: 17, area: 408 },
  { from: 17, area: 415 }
]
```

Output data

```
[
  { from: 17, area: 415, n: 2 },
  { from: 17, area: 408, n: 1 }
]
```

# Map-Reduce Jobs in Hadoop

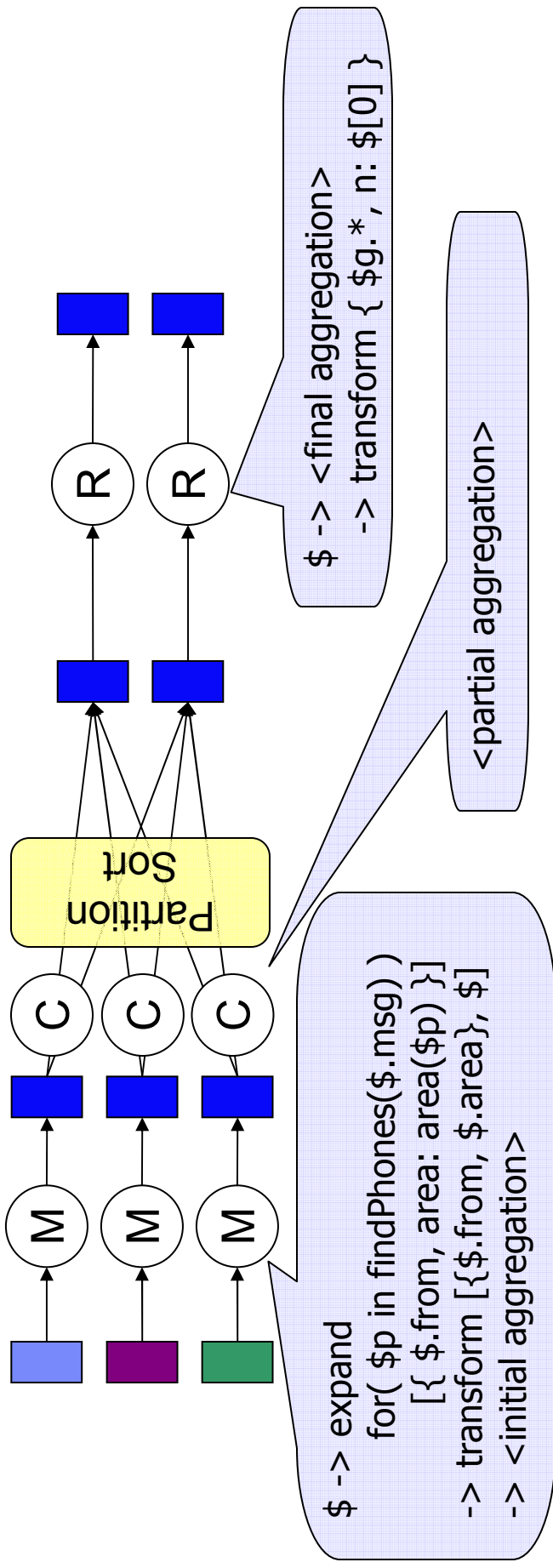
$[ V ] \rightarrow \mathbf{Map} \rightarrow [ K, V' ] \rightarrow \mathbf{Shuffle} \rightarrow [ K, [V'] ] \rightarrow \mathbf{Reduce} \rightarrow [ V'' ]$



```
mapReduce( {
  input : {uri: "hdfs:rawMsgs" },
  map   : fn($ ) ( $ -> expand for( $p in findPhones($.msg) )
                [ { $.from, area: area($p) } ]
                -> transform [ { $.from, $.area }, $ ] ),
  reduce : fn($g,$ ) ( $ -> transform { $g.*, n: count($) } ),
  output : { uri: "hdfs:out" } } )
```

# Exploiting Combiners with Algebraic Aggregates

$[ V ] \rightarrow \mathbf{Map} \rightarrow [ K, V' ] \rightarrow \mathbf{Shuffle} \rightarrow [ K, [ V' ] ] \rightarrow \mathbf{Reduce} \rightarrow [ V'' ]$

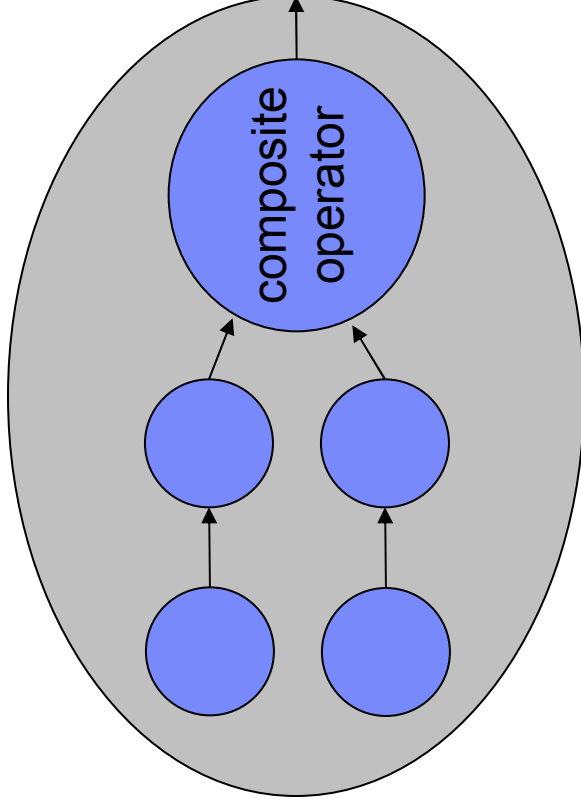


```

mrAggregate{
  input : {uri: "hdfs:rawMsgs" },
  map   : fn($)( $ -> expand for( $p in findPhones($.msg) )
            [ { $.from, area: area($p) } ]
            -> transform [ { $.from, $.area }, $ ] ),
  aggregate: fn($g,$) ( [ count($) ] )
  final : fn($g,$) ( $ -> transform { $g.*, n: $[0] } ),
  output : { uri: "hdfs:out" } }

```

# Composite Operators



## Examples:

- **Join**
  - Join two or more inputs on a key
  - Inner/outer/full
  - Multi-predicate, multi-way
- **Merge**
  - Concatenate all inputs in any order
- **Union, Intersect, Difference...**
- **User-defined function**

## CoGroup

Find users in each zipcode and the name of the city

```
$users = read("hdfs:users");
$cities = read("hbase:cities") -> expand unroll $.zips;

group $users by $zip = ($.users.zip),
    $cities by $zip = ($.cities.zips)
into {$zip, n: count($users), users: $users[*].name, city: singleton($cities[*].city) };
```

### Input data

```
users
[ { id: 12, name: "Joe Smith", zip: 94114 },
  { id: 17, name: "Ann Jones", zip: 95120 },
  { id: 19, name: "Alicia Fox", zip: 94114 } ]

cities
[ { city: "San Francisco", zips: [94112,94114], areas: [415] },
  { city: "San Jose", zips: [95120], areas: [408] } ]
```

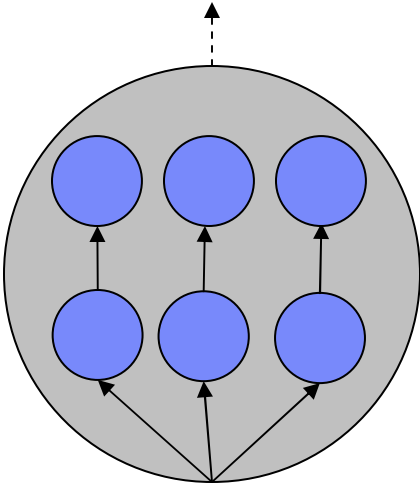
### after group

```
[ { zip: 94114,
  n: 2,
  users: [ "Joe Smith", "Alicia Fox" ],
  city: "San Francisco" },
  { zip: 95120,
  n: 1
  users: [ "Ann Jones" ],
  city: "San Jose" },
  { zip: 94112,
  n: 0,
  users: [],
  city: "San Francisco" } ]
```

## Multiple Outputs: tee

Send each input item to additional output pipe

```
read("hdfs:users")  
→ tee ( → filter $.zip == 94114  
        → write("hdfs:zip94114") )  
→ filter $.bday > date("1979-01-01")  
→ write("hdfs:over30");
```





## Rough Unix analogs of Jaql

Unix	Jaql
< filename, cat	read, merge
join	join
grep	filter
cut, paste, sed, tr	transform
sort	sort
head	top
uniq	distinct
> filename	write
tee	tee

Unix: stream of bytes / lines

Jaql: stream of JSON items  
more structure / types

## Other Map/Reduce Languages

- **Java**
  - The most basic way to write a map/reduce job in Hadoop
  - User implements a Mapper and Reducer interface and creates a JobConf
- **Hadoop Streaming**
  - Map and Reduce tasks implemented by external process in any language
  - Read/write data on stdin/stdout
- **Pig**
  - Nested relational model, but added “Map” for dynamic columns
  - From Yahoo. Hadoop subproject.
- **Hive**
  - Nested relational model with “Map” for dynamic columns and “Array” for lists
  - SQL-like syntax
  - From Facebook. Hadoop subproject.
- **Cascading**
  - A dataflow language, mostly relational.
  - Expressions are UDFs; Groovy extension
- **Proprietary and non-Hadoop**
  - Google Sawzall, Microsoft Dryad/LINQ, Microsoft Scope

## Current Roadmap

- **New syntax (presented here)**
  - About to be released to open source
- **Robustness**
  - Exception handling, ...
- **Features**
  - Programming/scripting language integration, schema, tooling, ...
- **Performance**
  - Query optimization, compact storage, runtime tuning, ...