Big Data Processing in Pharo

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Big Data Frameworks

Programming Model

To express large data processing jobs in terms of simple functions (e.g., Map/Reduce)

Parallel Execution Model

To execute in parallel computation on large amount of data





Why Big Data?

Google MapReduce

Index the WWW

More than 20 billion web pages (> 400 TB)

Do it fast

With 1000 machines: < 3 hours, with one machine, 4 months.

Keep it cheap

Use a lot of recycled hardware, but with robust error recovery



Map/Reduce by Example

Dataset

Rossi Abruzzo 1522706400

Verdi Toscana 1527976800

Bianchi Toscana 1525298400

Verdi Lombardia 1527976800

Gialli Toscana 1525298400

Rossi Abruzzo 1525298400

Gialli Veneto 1527976800

Bianchi Sardegna 1522706400

Gialli Toscana 1527976800

Bianchi Lombardia 1527976800

Bianchi Toscana 1522706400

Gialli Sicilia 1522706400

Gialli Sicilia 1527976800

Bianchi Sardegna 1527976800

KeyBy

(Rossi, Abruzzo ...)

(Verdi, Toscana ...)

(Bianchi, Toscana ...)

(Verdi, Lombardia ...)

(Gialli, Toscana ...)

(Rossi, Abruzzo ...)

(Gialli, Veneto ...)

(Bianchi, Sardegna ...)

(Gialli, Toscana ...)

(Bianchi, Lombardia

(Bianchi, Toscana ...)

(Gialli, Sicilia ...)

(Gialli, Sicilia ...)

(Bianchi, Sardegna ...)

Map

NIL

(Verdi, Toscana ...)

(Bianchi, Toscana ...)

NIL

(Gialli, Toscana ...)

NIL

NIL

NIL

(Gialli, Toscana ...)

 NIL

(Bianchi, Toscana ...)

NIL

NIL

NIL

GroupBy

(Verdi, Toscana)

(Bianchi, Toscana ...)

(Bianchi, Toscana ...)

(Gialli, Toscana ...)

(Gialli, Toscana ...)

Reduce

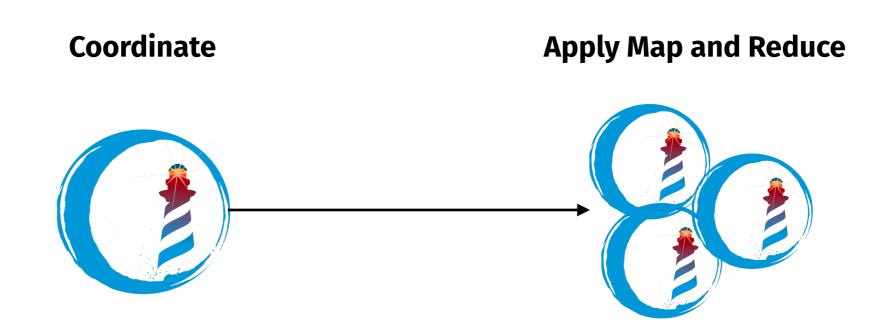
(Verdi, 1)

(Bianchi, 2)

(Gialli, 2)

A Map/Reduce application in Pharo

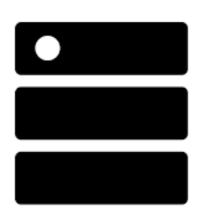
Port: a Map/Reduce Framework for Pharo



Deploying Port





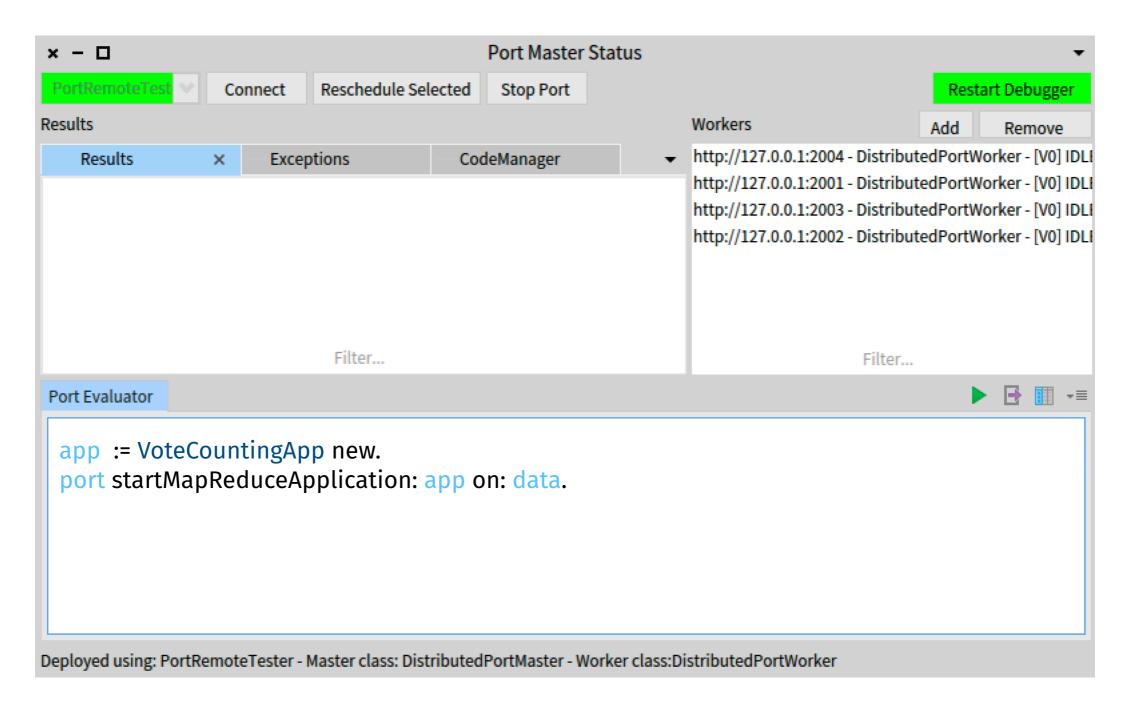


Local
Deploy on one machine using multiple cores.

StandaloneDeploy on multiple machines in the same network.

Yarn Mode
Deploy on a cluster using
Hadoop YARN.

Using Port



Handling Results

Asynchronous Execution

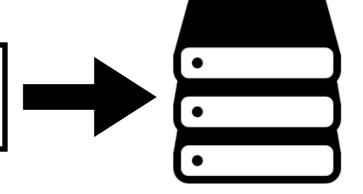
The application is executed asynchronously.

Asynchronous Result Handling

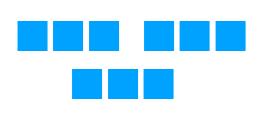
The result is handled asynchronously by the Map/Reduce Master

VoteCountingApp>>handleResult: resultPair

self storeToHDFS: resultPair



From Map/Reduce to Spark



Distributed Collection

Instead of executing Map/Reduce on a Dataset, you load the dataset in memory

aggregate:
filter:
map:
reduce:
groupBy:

Broader API

Not only execute map or reduce, but a broad set of methods applicable to the distributed collection



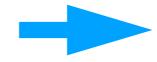
In-Memory operations

Avoid heavy storing of intermediate results by explicitly keep them in memory

The Spark-like API in Port

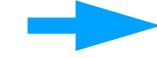
VoteCountingApp>>run

port startMapReduceApplication: app
 on: dataSet .



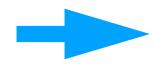
data := port distribute: dataSet.

VoteCountingApp>>map: aLine



result := data map: [...].

VoteCountingApp>>handleResult: res self storeToHDFS: res



result storeToHDFSPath: '...'.

The Polls Analyser in Port (V2)

result getCollection.
result storeToHDFSPath: '...'

DEMO

Uses of Port



Blockchain Analysis

In collaboration with Santiago at RMoD Lille. Full blockchain in 6 hours vs 2 days!



Genetic algorithms

In collaboration with Alexandre Bergel at Universidad de Chile

Towards Live Big Data Development Tools



Online Debugging

Debug the system as the bug is happening



Global View

Centralised debugging of the distributed system



Isolation

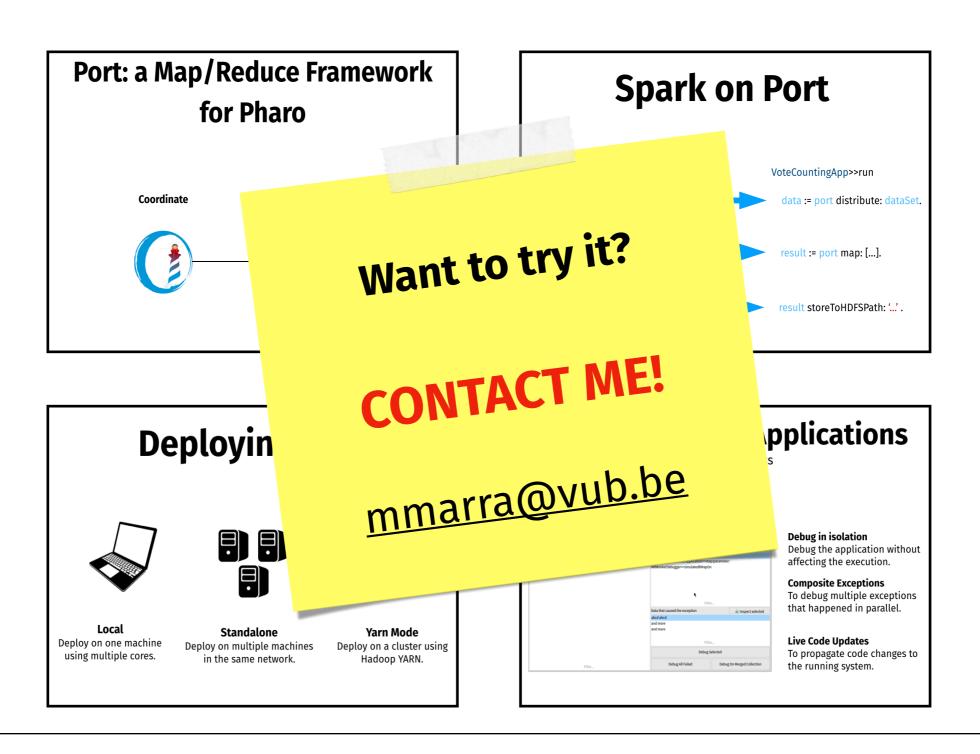
Debug the system without interfering with its execution



Update of the Running System

Deploy code-fixes without restarting the whole system

Conclusion



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