



DISPEL Introduction

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VERCE @ University of Liverpool, 3 September 2012

Outline

- Data Intensive
 - What is it?
 - Why use it?
- DISPEL
 - What is it?
 - Why design it?
 - Is it different?
- A simple example
- Streaming matters
- Summary and Conclusions



picture from Erica Salmon Cornish Coast Path; where I call home



Data-Intensive Thinking







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Gray's Laws of Data Engineering

Jim Gray:

- Scientific comp
- Need scale-out
- Take the analys
- Start with "20 c
- Go from "work



DATA-INTENSIVE SCIENTIFIC DISCOVERY

STOTED BY TONY HEY STEWART TANSLEY, AND KRISTIN TOLLS





From: Alex Szalay, JHU

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Defining "Data-Intensive"

Generally

- A computational task is data-intensive if you have to think hard about an aspect of data handling to make progress
 - distribution, permissions and rules of use, complexity, heterogeneity, rate of arrival, unstructured or changing structure, long tail of small and scattered instances, size of data, number of users
 - invariably in combination

Quantitatively

- The computation's Amdahl numbers are close to 1
 - CPU operations : bits transferred in or out of memory
 - 1000 CPU operations : 1 I/O operation
- Total volumes expensive to store
- Total requests/unit time hard to accommodate
- Data transport too slow or expensive



Data-Intensive Strategies 1

- Use commodity components and low power
 - So that you can afford a lot of them
 - Balanced for data-intensive work
 - Treat memory bandwidth as a scarce resource
- Data & computation as close together as possible
 - in the processor cache in fewest steps & not disrupted
- Work on small chunks of data
 - as small as logically possible
 - a column of a table
 - a row of a table
 - a file
 - data unbundled, in computational format & compressed
- Once data is close to a processor do all you can with it
 - multiple derivatives in one pass
 - pipelining
 - re-use of intermediate data, caching and forwarding
- Use catalogues and indexes to avoid revisiting large-volumes of data randomly



widely relevant

Data-Intensive Strategies 2

- Exploit very large scale parallelism and distribution
 - many subtasks at modest rate per task in large numbers
 - NOT tightly coupled parallelism!!!
 - distribution for availability, ownership & persistence
 - proximity to data sources or destinations for speed
 - good enough unbiased answers by careful sampling
- Replicate
 - for more parallelism and for durable persistence
- Most data WORM (Write Once Read Many)
 - or WORN (Write Once Read Never) automatically eliminate or clean up
- Updates local and mostly append (mostly non-Transactional)
- Coordination & Catalogue DBs
 - distributed shared structures
 - just enough synchronisation
- Fine-grained local protection & authorisation
- Statistical and quantised accounting



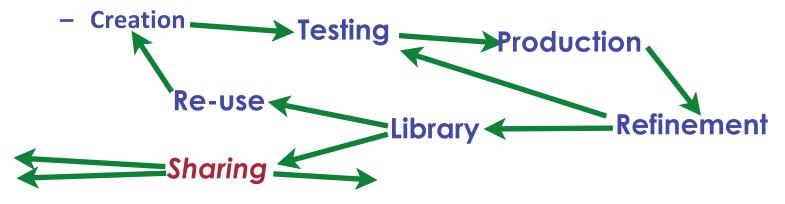
widely relevant

Data-Intensive Strategies 3

- High-level notations for describing methods /composing tasks
 - with well-developed optimised transformations before execution
 - query languages: SQL/AQL, (Xquery &SPARQL), ...
 - workflow languages: Kepler, Pegasus, DISPEL, ...
 - MapReduce: PigLatin, ZigZag, ...

widely relevant

- Providers + Community + User definition of (libraries of) tasks
 - your signal processing, geophysics & data-presentation steps
 - your existing code & preferred languages
- Support for the workflow lifetime: new research objects



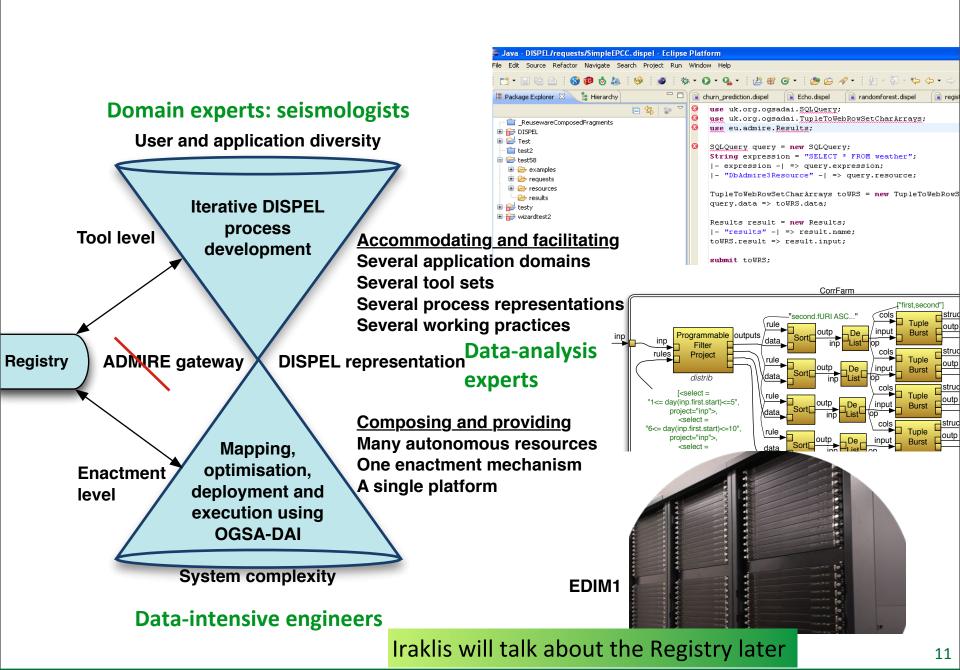


Data-Intensive Process Engineering Language

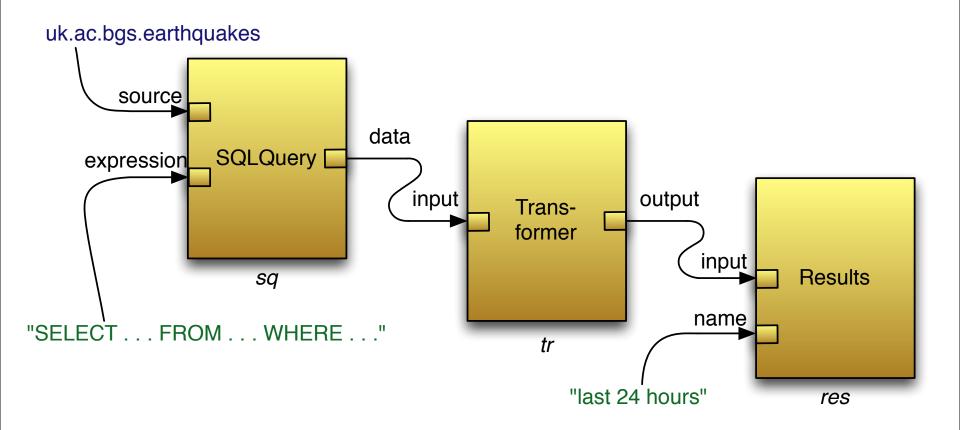
- A language for constructing data-flow graphs
 - Nodes are processing elements
 - Arcs are data-flow paths
- A language for generating data-flow patterns
 - Functions hide detail of graphs
 - Functions generate graphs
- A language for discussing data-flow engineering
 - Designed to be read and written by humans
 - As well as by programs
 - Supports validation and optimisation

designed to encourage data-intensive thinking





A simple DISPEL graph





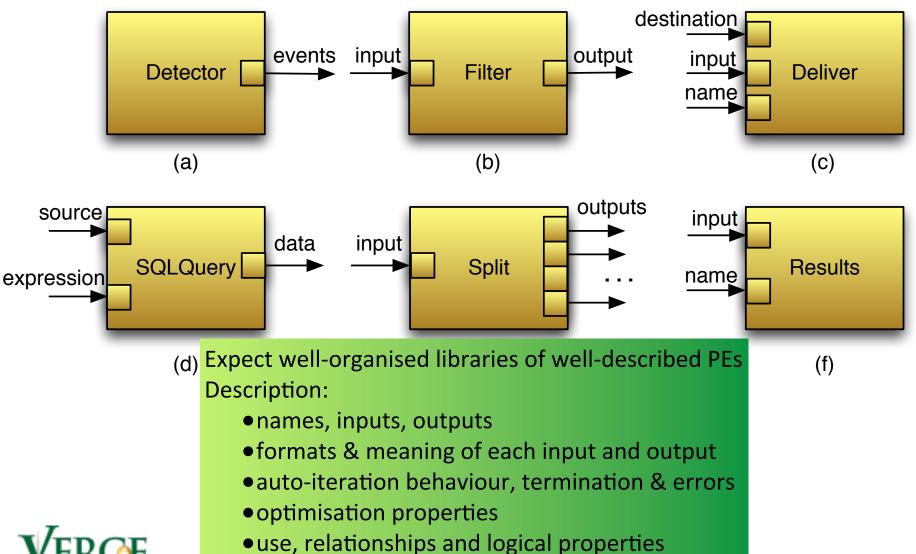
The DISPEL to Generate it

```
//set working context
package book.examples.seismology {
                                                     //import PE SQLQuery
  use dispel.db.SQLQuery;
                                                     //import PE Transform
  use book.examples.seismo.Transform;
                                                     //import PE Results
  use dispel.lang.Results;
                                                     // new instance of SQLQuery
  SQLQuery sq = new SQLQuery;
                                                     // new instance of Transform
  Transform tr = new Transform;
                                                     // new instance of Results
  Results res = new Results;
                                                     // set up data flow from sq to tr
  sq.data => tr.input;
                                                     // set up data flow from tr to res
  tr.output => res.input;
                                                     // URI of source of data
  |- "uk.ac.bgs.earthquakes" -| => sq.source;
  |- "SELECT ... FROM ... WHERE ..." -| => sq.expression; //query gets traces
                                                     //name of results for user
  |- "last 24 hours" -| => res.name;
                                                     // submit for enactment
  submit res;
```



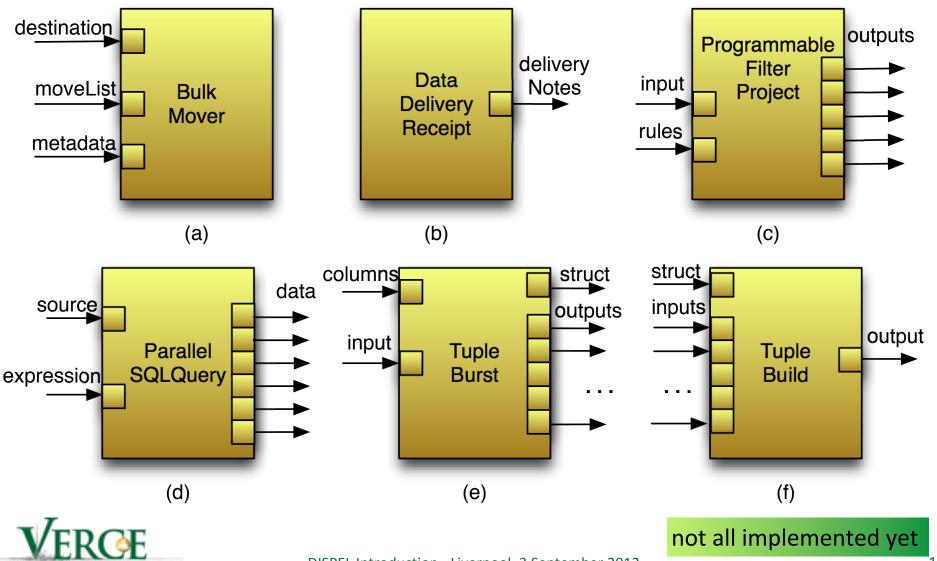
Paul will talk about this next

Processing Elements

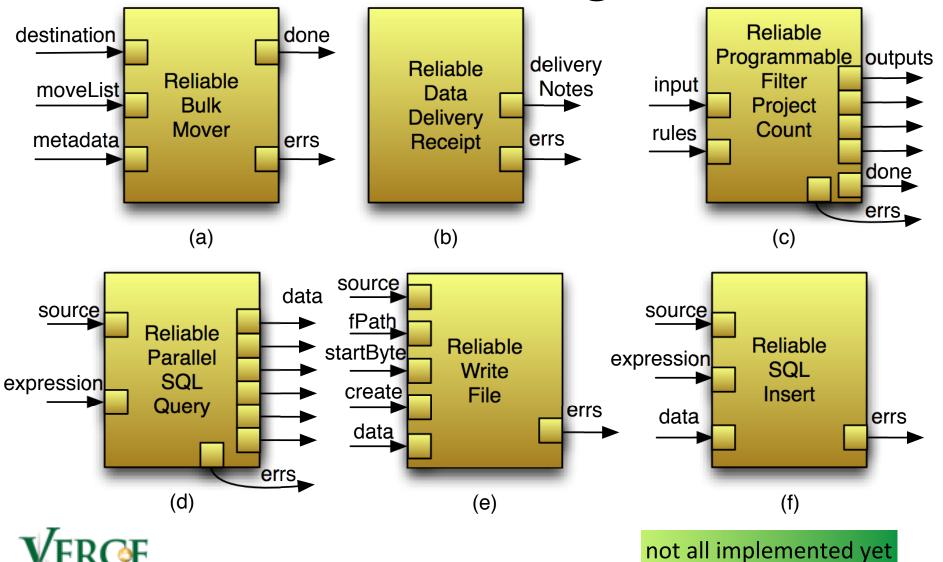


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Enhanced Processing Elements

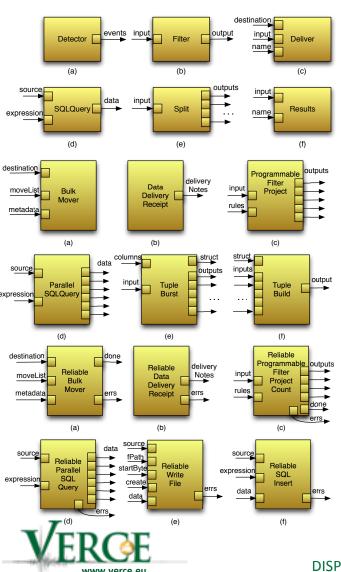


Advanced Processing Elements



112

Processing Element Libraries



- Good libraries are needed
 - -Generic libraries Amy builds these
 - -Domain specific libraries
- KNMI building VERCE libraries
 - -Hard work today

Alessandro and Luca will talk about these and this work tomorrow

- Need harnesses & tools
 - -To quickly wrap existing libraries
 - -To quickly deploy new algorithms

Functions

- Algorithms to generate graphs
 - parametric variation
 - patterns
 - parameters
 - subgraphs

Expect well-organised libraries of well-described Functions Description:

- names, type signature, description, ...
- Abstraction and Optimisation
 - smart methods for common patterns
 - hiding pattern implementation for stability
 - late evaluation would permit contextual



we return to functions after Paul's talk

Enactment Model

- 1. DISPEL language processing
 - 1.1. Validation & import from registry
 - 1.2. Format & meaning mis-match handling
 - 1.3. Interpretation to generate graph
- 2. Graph optimisation & mapping
 - 2.1. Re-ordering & parallelisation
 - 2.2. Identification of where to do the work
 - 2.3. Selection of PE implementations & instances
 - 2.4. Partitioning into co-located subgraphs
- 3. Deployment & initialisation
- 4. Execution, Monitoring & Clean up

Preparing to do the data processing

Doing the data processing

Amy will talk about this later



DAGMan



























Ian J. Taylor, Ewa Deelman, Dennis B. Gannon, and Matthew Shields. **Workflows for e-Science: Scientific Workflows for Grids.** Springer London, 2007.



DISPEL is Different 1

- Spanning Distributed Independent Hosts
 - Fragments of one workflow can run in different regimes
 - Different security models
 - Different file systems
 - Different DBMS
 - Different Operating Systems
 - Different DISPEL implementations
- Agnostic about Size & Scale
 - Processing Elements of any size
 - Data values in streams of any size
 - Streams of any length
 - Graphs of any size

Can process data much larger than local storage



DISPEL is Different 2

- Patterns & Pattern Composition
 - Functions define & generate patterns
 - Higher-order functions compose patterns
 - Functions can be refined to optimise
- Component-Description Driven
 - Rich description of components
 - Capturing logical properties
 - Collecting component-builders' hints
- Restricted language for workflow longevity
 - Only hints and no WF-definition time concrete mappings
 - Late mapping potentially permits optimisation, for the system as it is at execution time usually much different from definition time!



Streaming



Why Stream

- Couple data-processing steps
 - locally or across networks
- Handle continuous or very large data
 - incrementally by auto-iteration on units
 - for any scale of unit
- Provide an opportunity to cancel
 - as soon as partial results show problems
- Accelerated processing
 - overlapping processing & transmission steps
 - potentially multiple steps in same cache





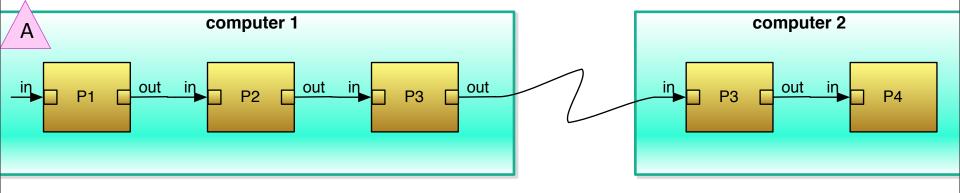
A Single Stream

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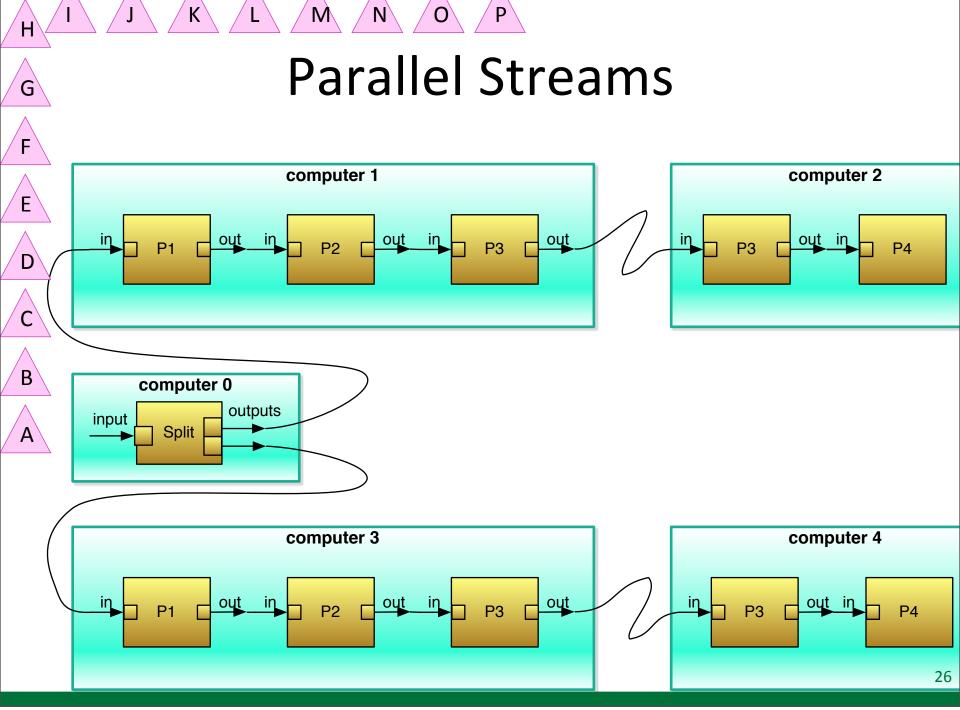
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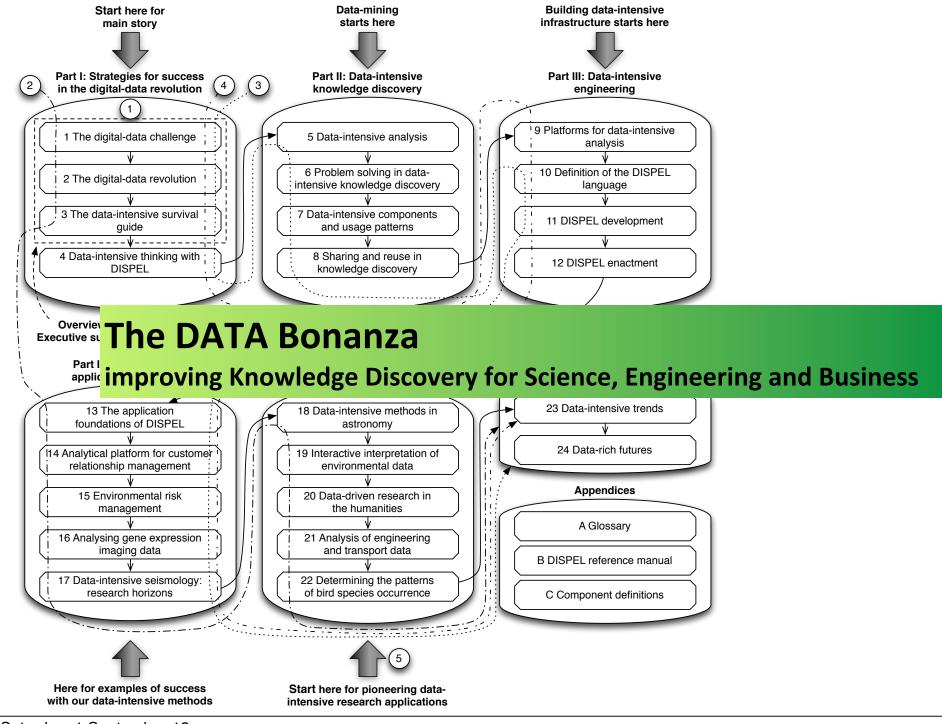






Summary and Conclusions





Summary

- DISPEL is an experimental data-intensive language
 - draws on workflows & database query internals
 - auto-iteration over values flowing through connections
 - agnostic about value sizes implementation challenge
 - controlled access to system information
 - optimisation based on description & operation
 - distributed termination protocol
- Several years of experience
 - seven different application domains
- Differences
 - functional pattern handling
 - multi-scale streams
 - restricted information to permit platform evolution
- Status
 - two implementations: to OGSA-DAI & to Java
 - much still to do to fully explore the ideas



Today's Programme

- Introduction Malcolm Atkinson √
- DISPEL language Paul Martin
- DISPEL enactment Amy Krause
- DISPEL functions Malcolm Atkinson
- Registry Iraklis Klampanos



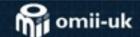


www.verce.eu

research.nesc.ac.uk/node/828

www.ogsadai.org.uk

epcc



www.omii.ac.uk









Picture composition by Luke Humphry based on prior art by Frans Hals