

dispel4py in detail

Amy Krause - EPCC Rosa Filgueira- School of Informatics

University of Edinburgh

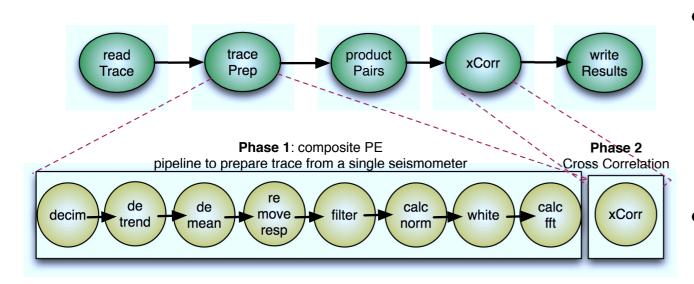
Outline

- Installation and links
- dispel4py workflows
 - Seismology
 - Astrophysics
 - Social Computing
- Graph examples
- Composite PEs
- Chains
- Groupings
- Mappings to execution platforms

Installation & links

- This is all you need:
 - pip install dispel4py
- GitHub: https://github.com/dispel4py/dispel4py
- Documentation: http://www2.epcc.ed.ac.uk/
 ~amrey/dispel4py
- **Exercises**: http://effort.is.ed.ac.uk/dispel4py_exercises.tar

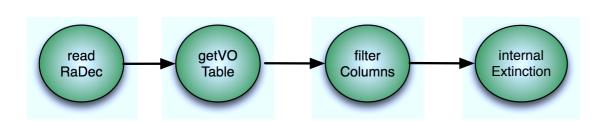
dispel4py workflows-Seismology



- Phase 1- Preprocess: Time series data (traces) from seismic stations are preprocessed in parallel
- Phase 2: Cross-Correlation:
 Pairs all of the stations and calculates the cross-correlation for each pair (complexity O(n2)).
- Input data: 1000 stations as input data (150 MB)
- Output data: 499,500 crosscorrelations (39GB)
- 11th IEEE eScience 2015 paper

dispel4py: An Agile Framework for Data-Intensive eScience

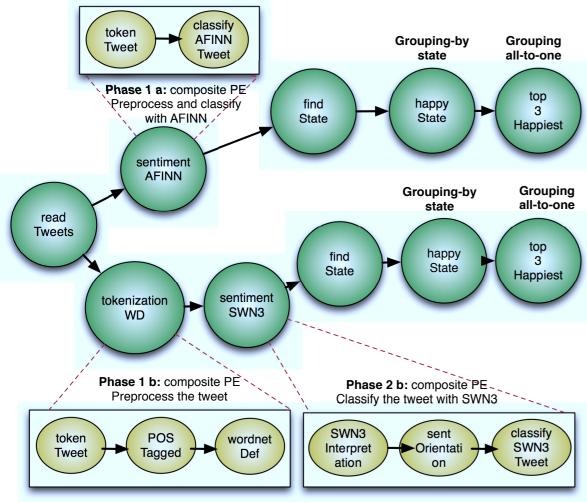
dispel4py workflows-Astrophysics



- Calculates the Internal Extinction of the Galaxies from the AMIGA catalogue
- This property represents the dust extinction within the galaxies and is a correction coefficient needed to calculate the optical luminosity of a galaxy.
- The first PE reads an input file (coordinates.txt size 19KB) that contains the right ascension (Ra) and declination (Dec) values for 1051 galaxies.
- The second PE queries a VO service for each galaxy in the input file using the Ra and Dec values.
- The results of these queries are filtered by filterColumns PE. It selects:
 - morphological type (Mtype)
 - apparent flattening (logr25)
- Their internal extinction is calculated by the internal extinction PE.
- 11th IEEE eScience 2015 paper

dispel4py: An Agile Framework for Data-Intensive eScience

dispel4py workflows-Social Computing

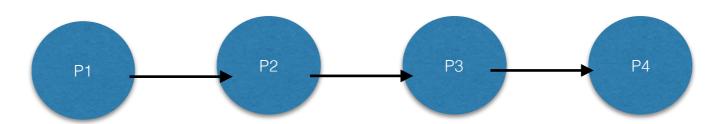


11th IEEE eScience 2015 paper

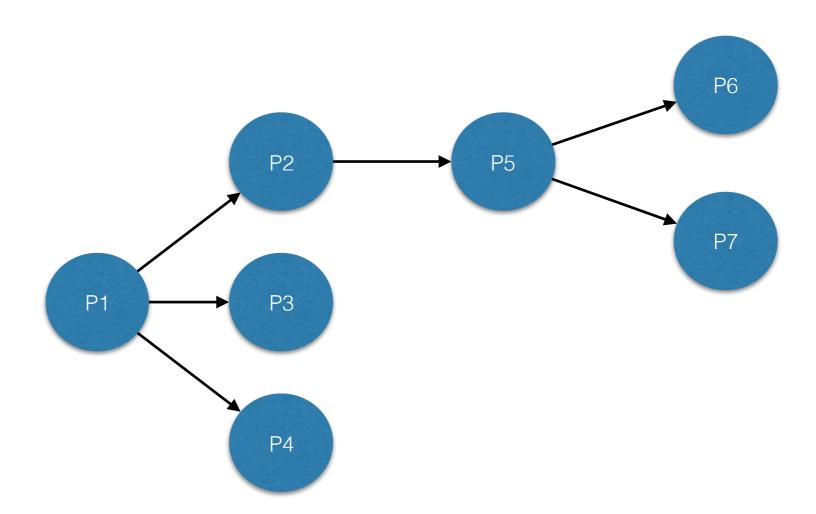
dispel4py: An Agile Framework for Data-Intensive eScience

- Two basic sentiment analyses by applying two lexicons:
 - **AFINN** (2477 English words)
 - SentiWordNet (155,287 English words and 117,659 synsets)
- After tweets findState PE, which searches the US state from which the tweet originated
- The HappyState PE applies a grouping by based on the state and aggregates the sentiment scores of tweets from the same state
- The last PE applies all-to-one grouping and determines which are the top three happiest states.
- 126,826 tweets (500MB)

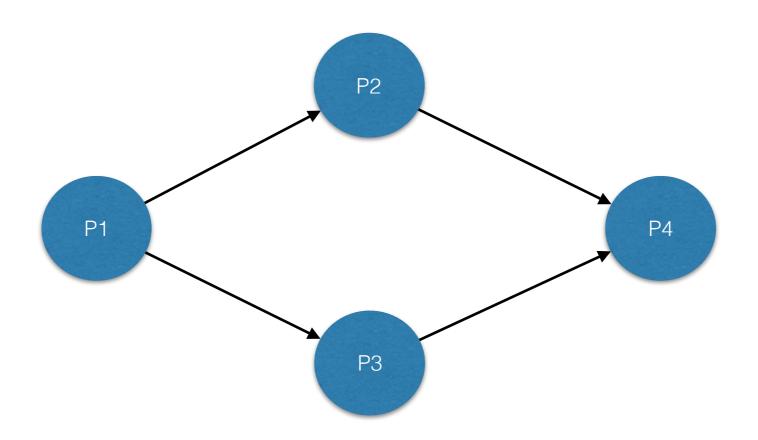
Pipeline



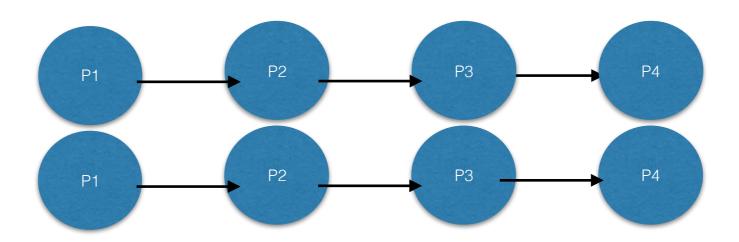
Tree



Split and Merge

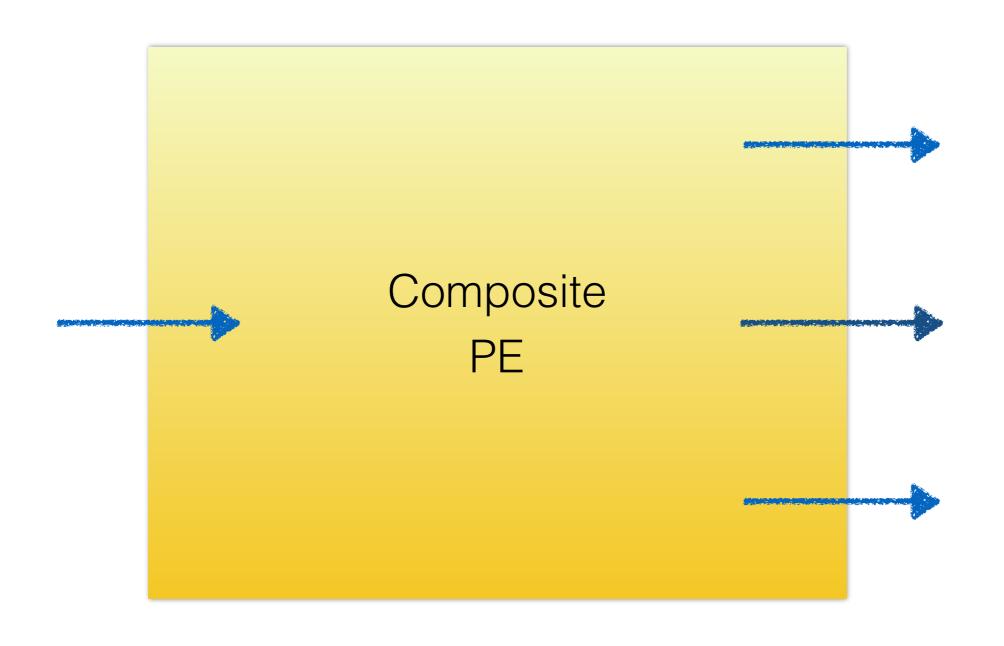


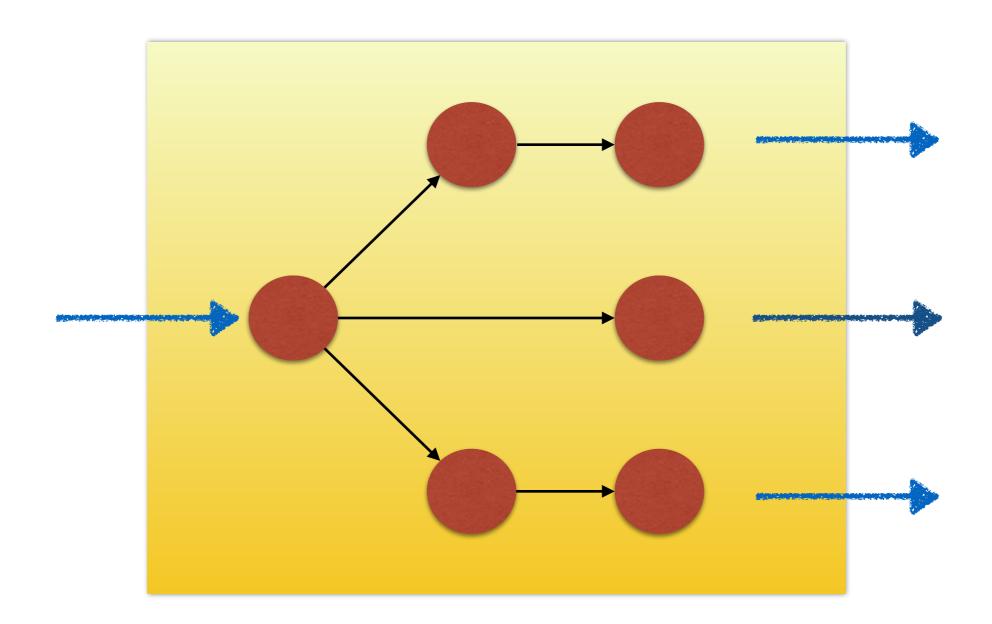
Unconnected

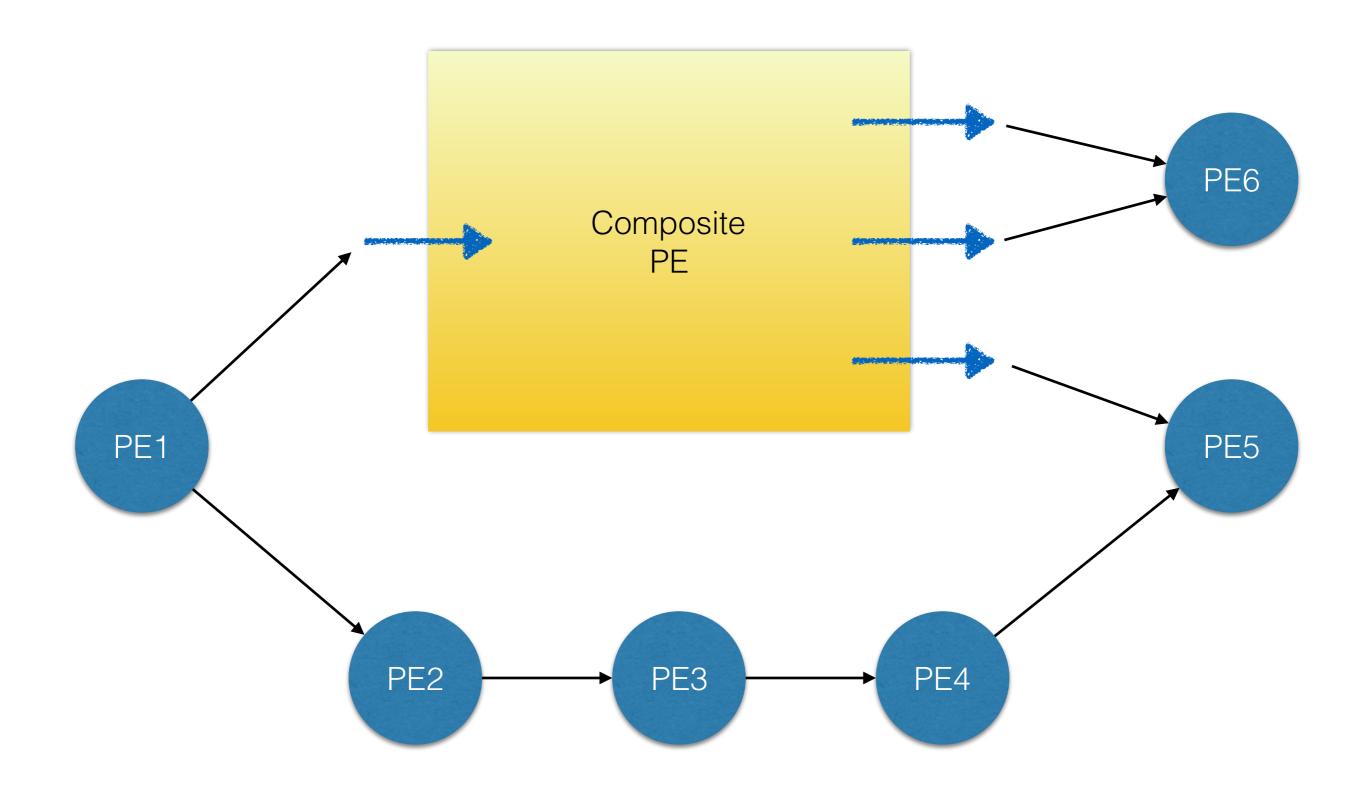


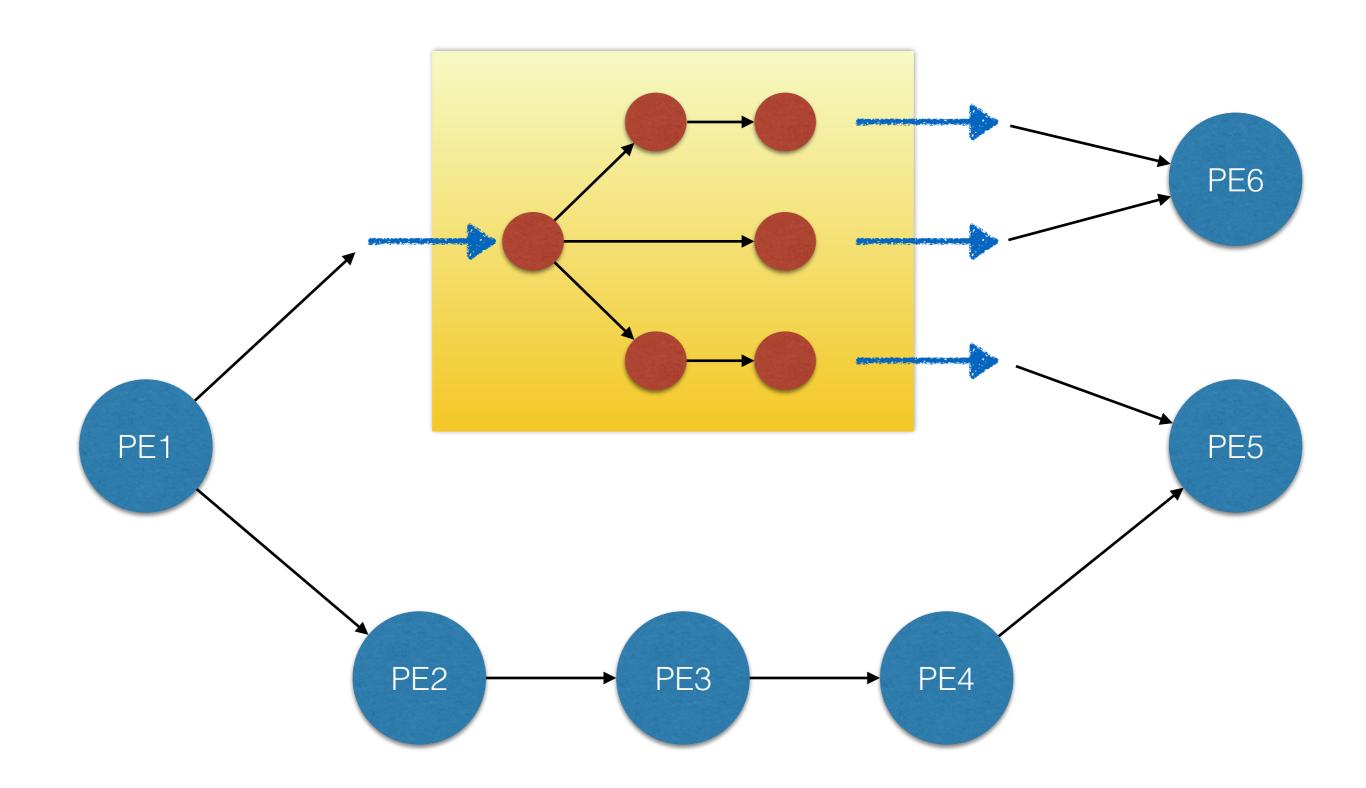
Composite PEs

- A composite PE is a nested graph
- Looks like a PE but contains other PEs
- Hides the complexity of an underlying process
- When creating a graph, a composite PE is treated like any other PE



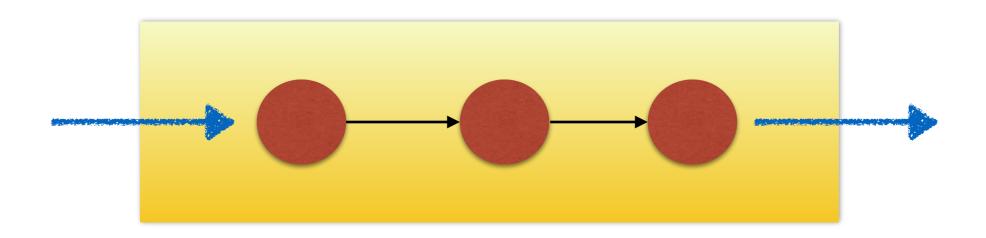




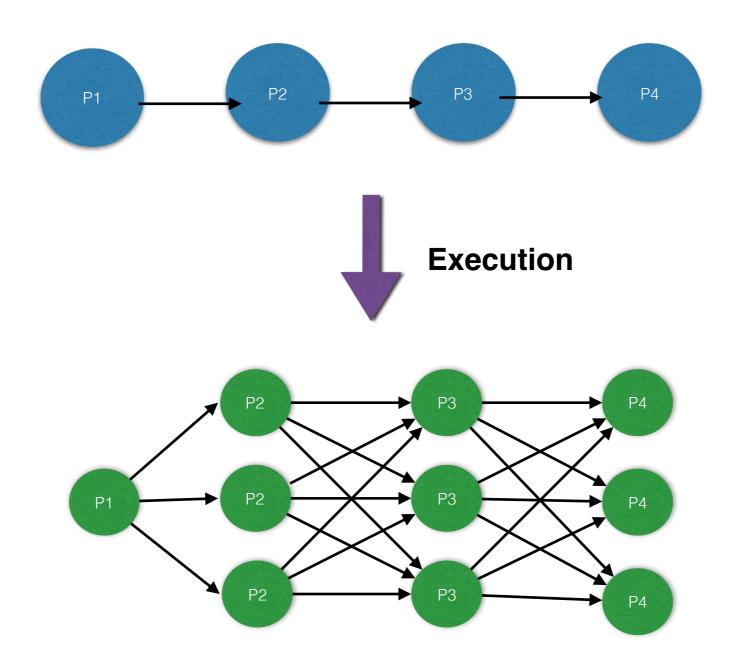


```
def add_value(num, value):
    num += value
    return num
def subtract_value(num, value):
    num -= value
    return num
pipeline = [ add_value, subtract_value, ... ]
```

compositePE = create_iterative_chain(pipeline)



Executing graphs



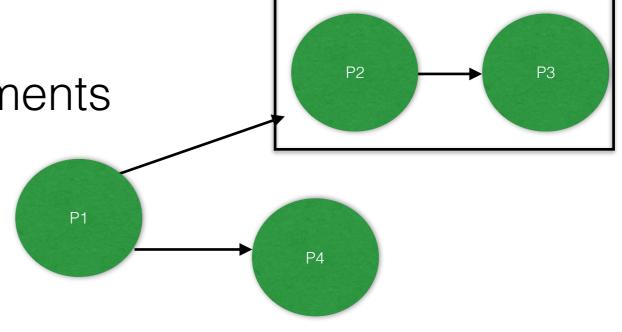
Instance

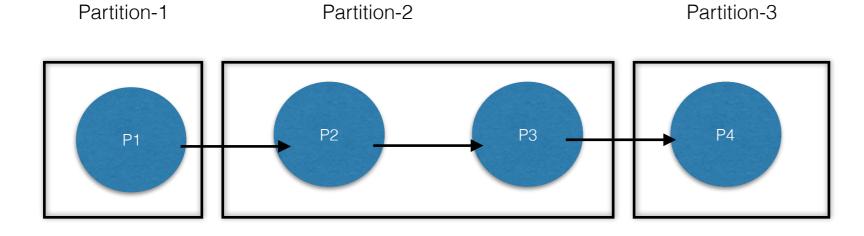
Partitions

Run several PEs in a single process

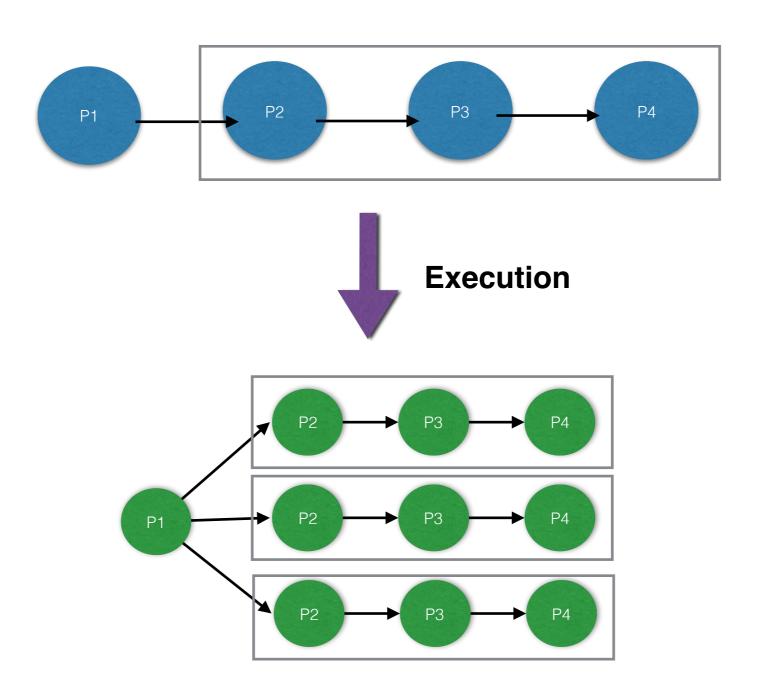
User defined

Applies to parallel environments





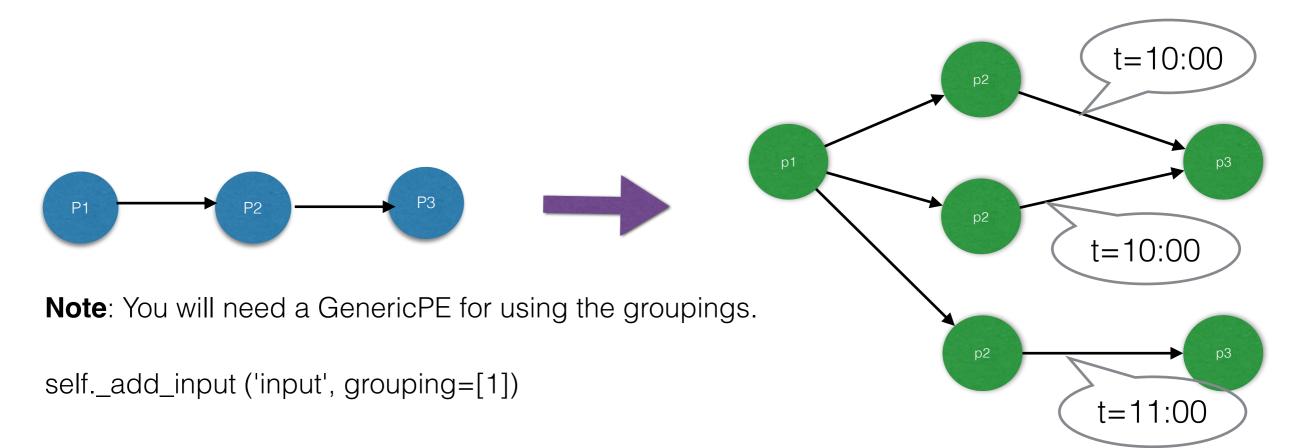
Partitioned Pipeline



Groupings

"Grouping by" a feature (MapReduce)

All data items that satisfy the same feature are guaranteed to be delivered to the same **instance** of a PE



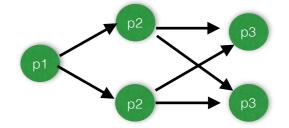
Other groupings

One-To-All



P3 - grouping "all":

P2 instances send copies of their output data to **all** the connected instances

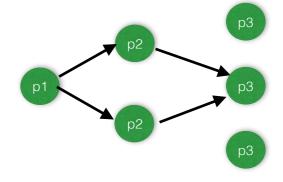


Global



P3 - grouping "global":

All the instances of P2 send all the data to **one** instance of P3



Mappings

Simple process

Sequential mapping for local testing

Multi process

- Parallelism based on processes, using Python's multiprocessing library
- Shared memory

MPI

- Distributed Memory, message-passing parallel programming model
- Practical, portable, efficient, flexible and stable
- Many HPC centres support it, and it has been widely used in the HPC community

STORM:

- Distributed Real-Time computation System
- Fault-tolerant and scalable

Running graphs

Sequential mapping

>> dispel4py simple <name_dispy_graph> <-f input_file in JSON format>

E.g. dispel4py simple dispel4py.examples.graph_testing.pipeline_test Multi-process mapping

>> dispel4py multi <name_dispy_graph> -n <number mpi_processes> <-f input_file in JSON format> <-s>

E.g: dispel4py multi dispel4py.examples.graph_testing.pipeline_test -n 6

MPI mapping

>> mpiexec -n <number mpi_processes> dispel4py mpi <name_dispy_graph> <-f input_file in JSON format> <-s>

E.g: mpiexec -n 6 dispel4py mpi dispel4py.examples.graph_testing.pipeline_test

STORM mapping

>> python storm_submission.py <name_dispy_graph> <-m mode: [remote, local, create]>

E.g: dispel4py storm dispel4py.examples.graph_testing.pipeline_test -m remote

Extra useful dispel4py information

- dispel4py commands:
 - dispel4py -h
- -h, --help show this help message and exit
- -a attribute, --attr attribute >> name of graph variable in the module
- -f inputfile, --file inputfile >> file containing input dataset in JSON format
- -d inputdata, --data inputdata >> input dataset in JSON format
- -i iterations, --iter iterations >> number of iterations

Extra useful dispel4py information

- inputs to the workflow:
 - dispel4py simple workflow.py -d '{"PE NAME CLASS" : [{"input" : "Hello World"}]}"
 - dispel4py simple workflow.py -d '{"PE NAME CLASS" : [{"input" : "coordinates.txt"}]}'
 - dispel4py simple workflow.py -f coordinates.txt

Thanks!