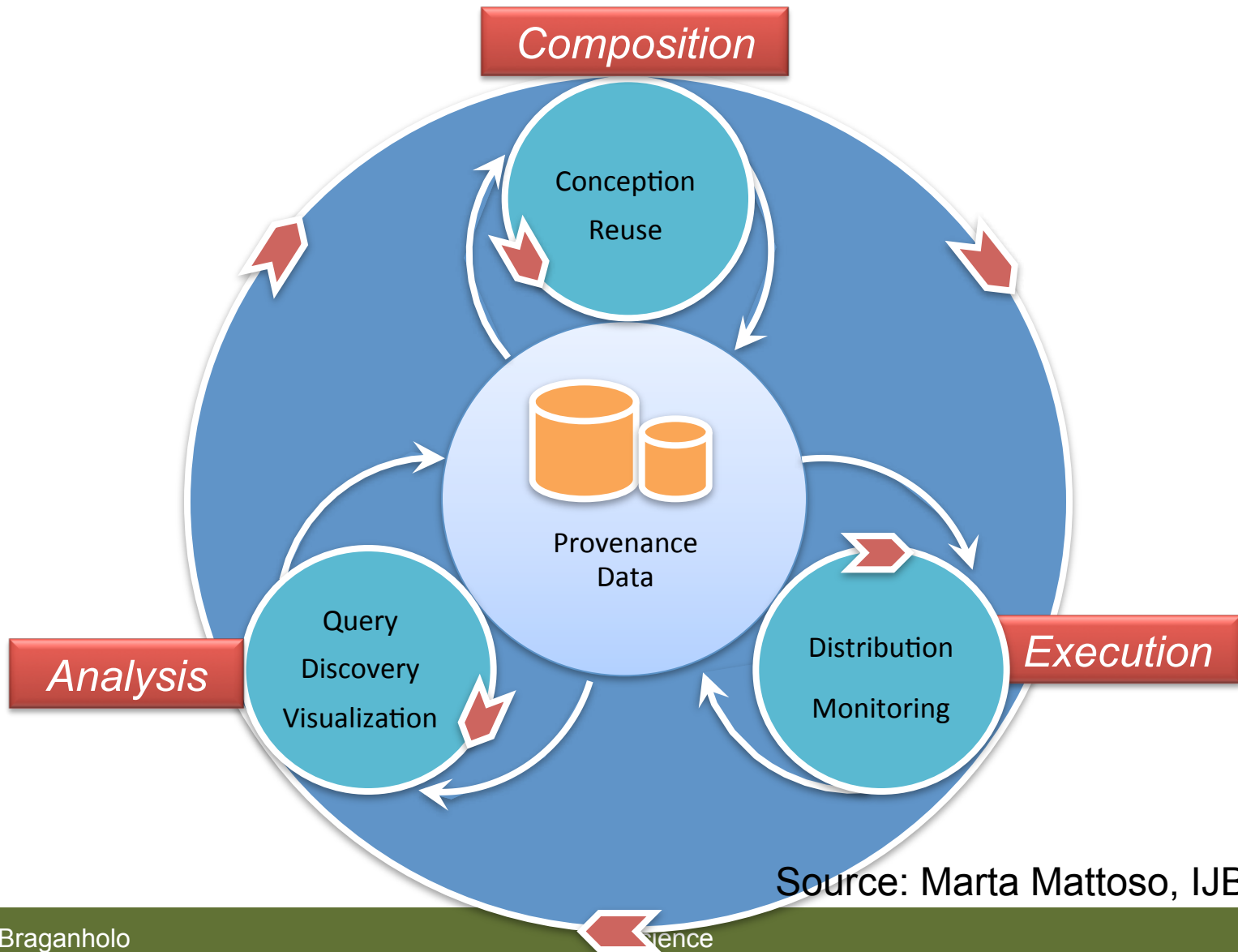


Scientific Experiments as Workflows and Scripts



The experiment life cycle



Source: Marta Mattoso, IJBPM, 2010

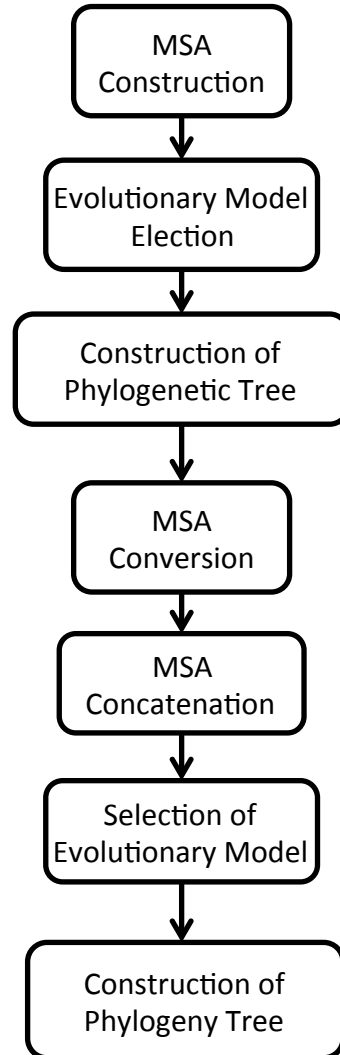
Agenda

- Abstract Representation of Scientific Experiments
- Workflows
- Scripts
- Black Boxes X White Boxes
- Workflow Management Systems
- Provenance Management Systems for Scripts

Composition: Conceiving Scientific Experiments

- Scientists usually design an experiment using a **high abstraction level representation** that is later mapped into a workflow or script

Phylogeny Analysis Experiment (Abstract Workflow)



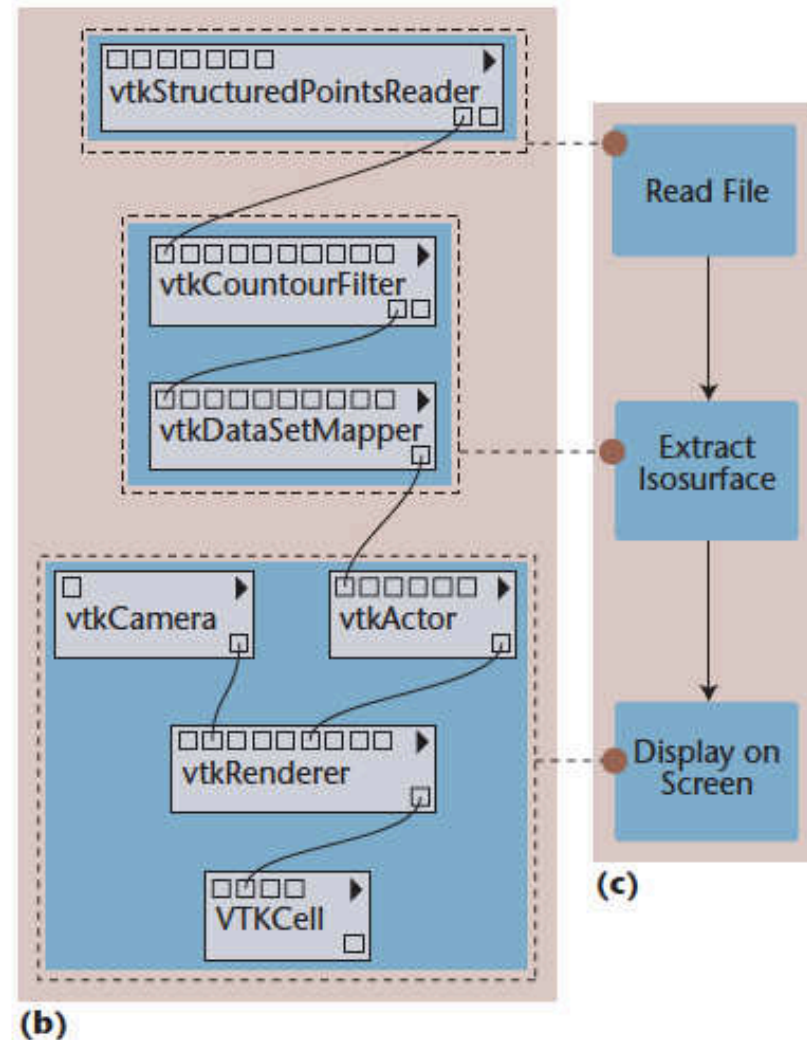
Abstract x Concrete

- The **abstract** workflow is later mapped into a **concrete** workflow or script


```

1  import vtk
2
3  data = vtk.vtkStructuredPointsReader()
4  data.SetFileName("../.../examples/data/head.120.vtk")
5
6  contour = vtk.vtkContourFilter()
7  contour.SetInput(0, data.GetOutput())
8  contour.SetValue(0, 67)
9
10 mapper = vtk.vtkPolyDataMapper()
11 mapper.SetInput(contour.GetOutput())
12 mapper.ScalarVisibilityOff()
13
14 actor = vtk.vtkActor()
15 actor.SetMapper(mapper)
16
17 cam = vtk.vtkCamera()
18 cam.SetViewUp(0, 0, -1)
19 cam.SetPosition(745, -453, 369)
20 cam.SetFocalPoint(135, 135, 150)
21 cam.ComputeViewPlaneNormal()
22
23 ren = vtk.vtkRenderer()
24 ren.AddActor(actor)
25 ren.SetActiveCamera(cam)
26 ren.ResetCamera()
27
28 renwin = vtk.vtkRenderWindow()
29 renwin.AddRenderer(ren)
30
31 style = vtk.vtkInteractorStyleTrackballCamera()
32 iren = vtk.vtkRenderWindowInteractor()
33 iren.SetRenderWindow(renwin)
34 iren.SetInteractorStyle(style)
35 iren.Initialize()
36 iren.Start()

```



Source: Freire et al., 2008. Provenance for Computational Tasks: A Survey.

Scientific Workflow

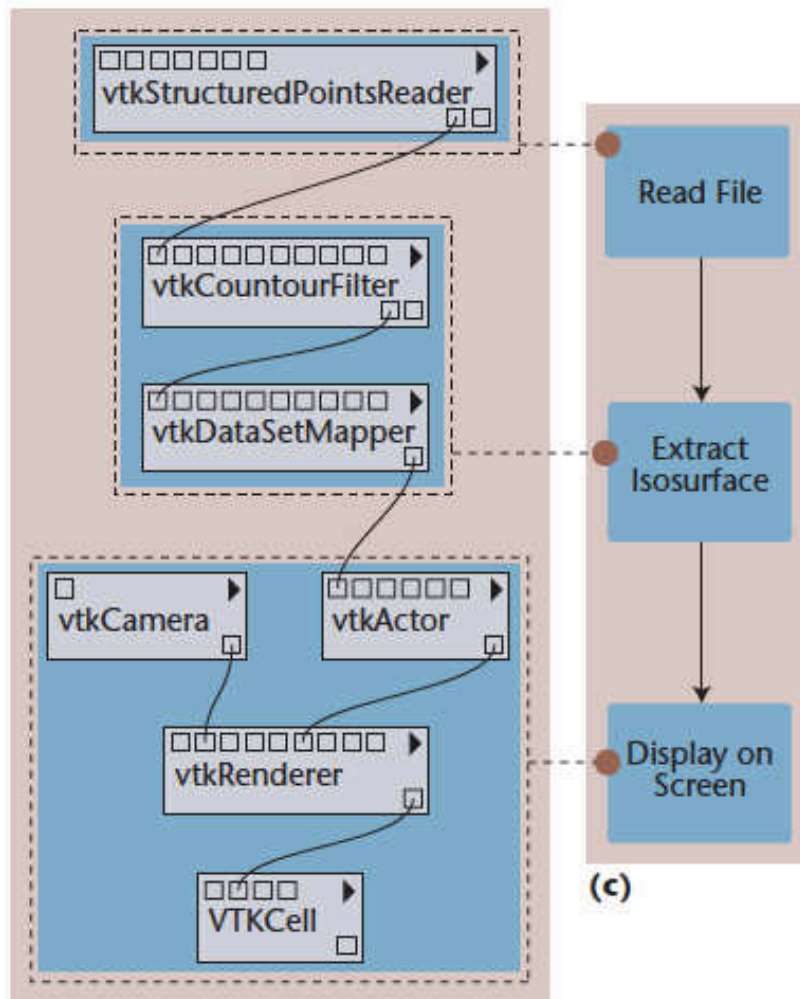
- A scientific workflow is a **chain of activities** organized in the form of a **data flow**

Data Flow

- In a data flow, the **execution is guided by the data**
- As soon as all the input data of an activity is available, it starts executing

Example

- Activities
vtkStructuredPointsReader and **vtkCamera** do not depend on other activities data, so they can start executing right away



Source: Freire et al., 2008. Provenance for Computational Tasks: A Survey.

Script

- Definition is controversial
- One of the most accepted definitions is that a script language is a “**programming language that does not require an explicit compilation step**”
- In other words, scripts are usually written in Languages that are interpreted instead of compiled
- Examples: Python, R, MatLab, etc.

Script

- Execution follows a **control flow** instead of a data flow
 - Commands explicitly define the execution order

```

1| DRY_RUN = ...
2|
3| def process(number):
4|     while number >= 10:
5|         new_number, str_number = 0, str(number)
6|         for char in str_number:
7|             new_number += int(char) ** 2
8|         number = new_number
9|     return number
10|
11| def show(number):
12|     if number not in (1, 7):
13|         return "unhappy number"
14|     return "happy number"
15|
16| n = 2 ** 4000
17| final = process(n)
18| if DRY_RUN:
19|     final = 7
20| print(show(final))

```

Source: Pimentel et al., 2016. Fine-grained Provenance Collection over Scripts Through Program Slicing

Running an Experiment

- A workflow or script is just part of an experiment
- In order to prove or refute an hypothesis, it is usually necessary to run the workflow or script several times, varying inputs, parameters and programs
- Each of those runs is called a **trial** of the experiment

New experiment!

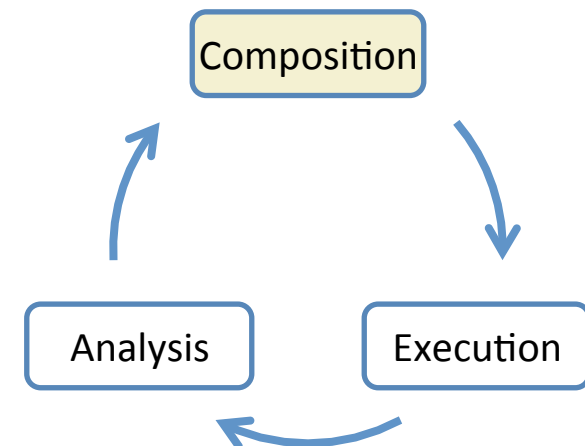
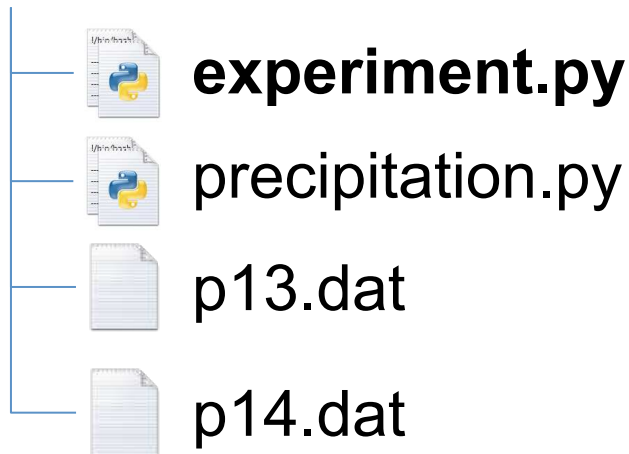
Could you check if the precipitation
of Rio de Janeiro remains constant
across years?



1st Iteration

- $H \downarrow 1$: “The precipitation for each month remains constant across years”

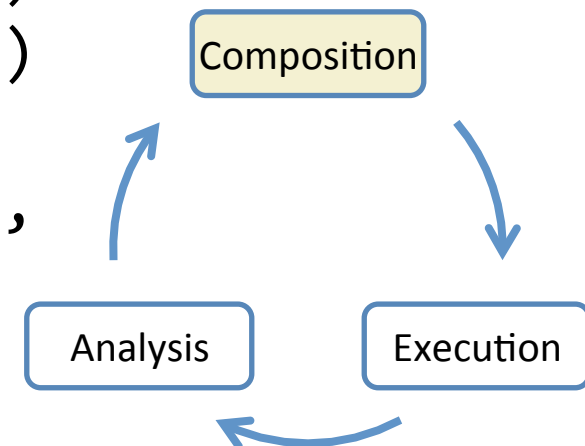
 Project Data: 2013, 2014 [BDMEP]



```

1| import numpy as np
2| from precipitation import read, sum_by_month
3| from precipitation import create_bargraph
4|
5| months = np.arange(12) + 1
6|
7| d13, d14 = read("p13.dat"), read("p14.dat")
8|
9| prec13 = sum_by_month(d13, months)
10| prec14 = sum_by_month(d14, months)
11|
12| create_bargraph("out.png", months,
13|     ["2013", "2014"],
14|     prec13, prec14)

```



Trial

\$ now run -e Tracker experiment.py



Project



experiment.py



precipitation.py



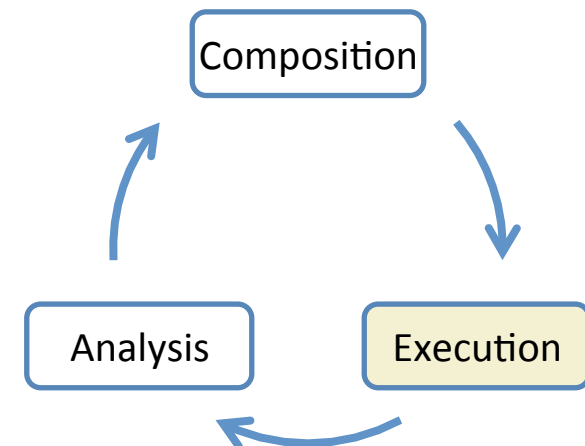
p13.dat

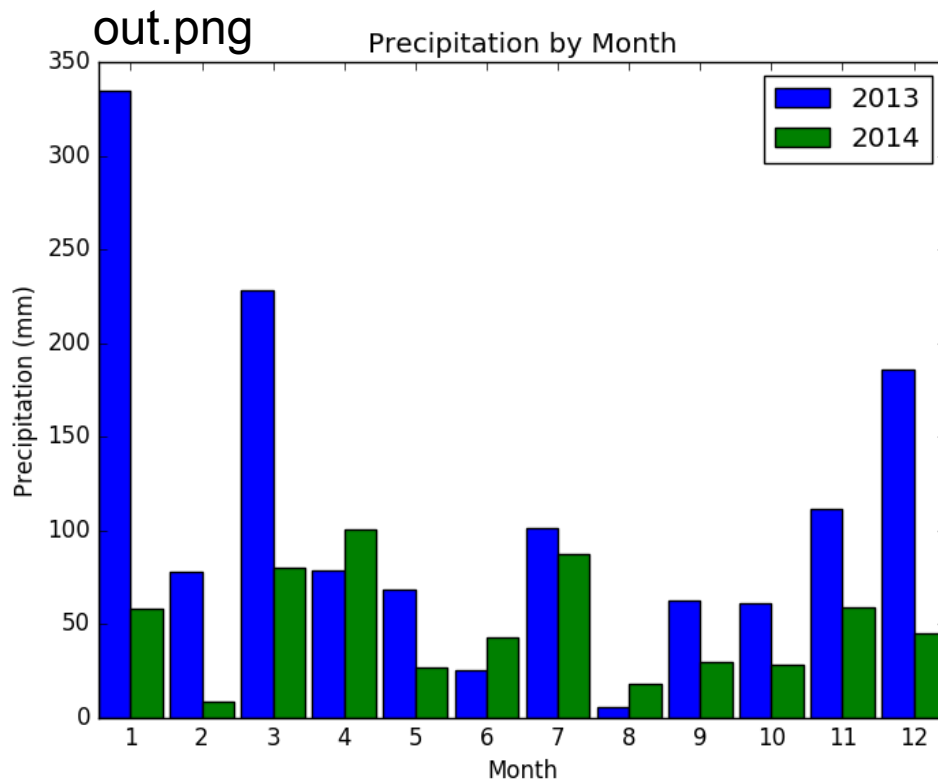


p14.dat



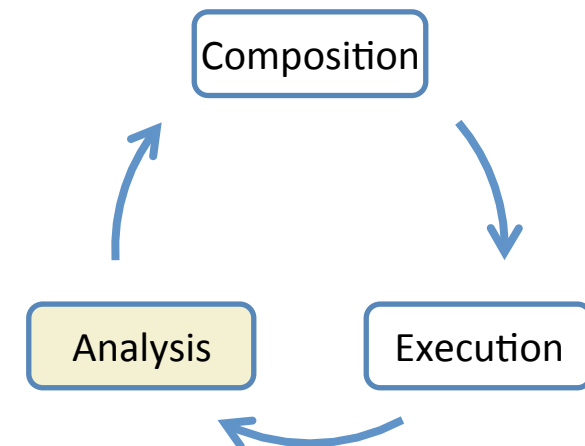
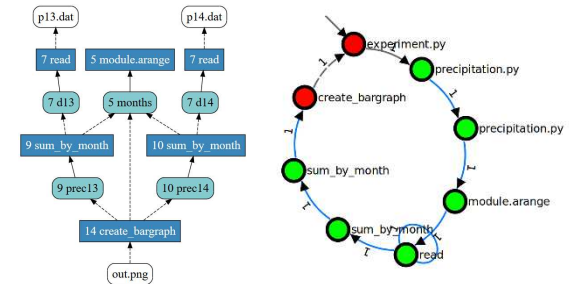
out.png





Conclusion:
“Drought in 2014”

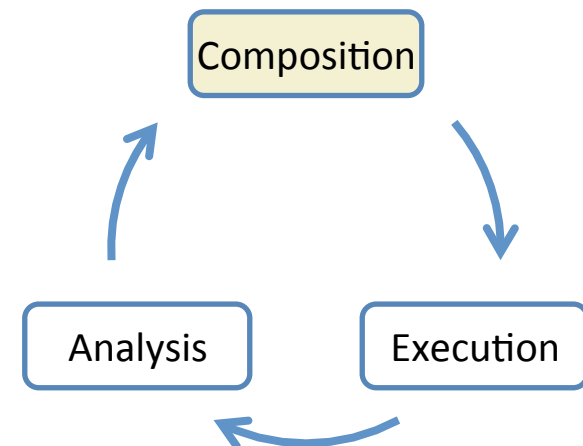
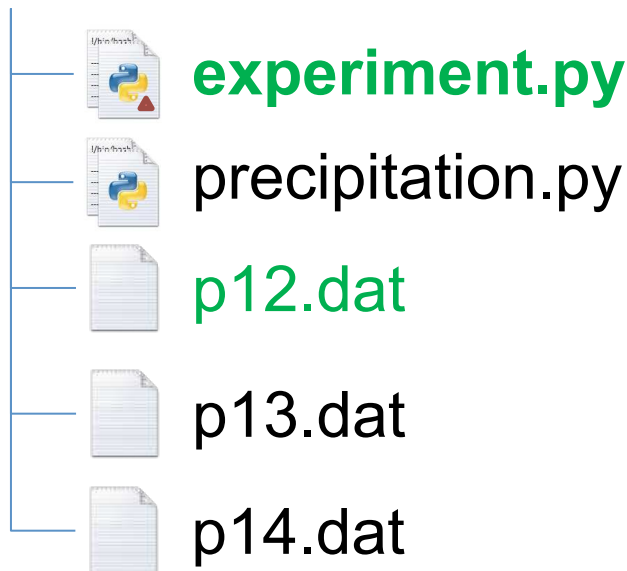
SELECT ...



2nd Iteration

- $H \downarrow 2$: “The precipitation for each month remains constant across years **if there is no drought**”

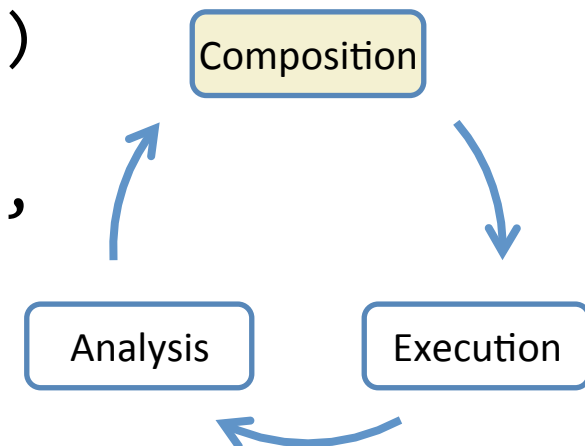
 Project Data: **2012**, 2013, 2014 [BDMEP]



```

1| import numpy as np
2| from precipitation import read, sum_by_month
3| from precipitation import create_bargraph
4|
5| months = np.arange(12) + 1
6| d12 = read("p12.dat")
7| d13, d14 = read("p13.dat"), read("p14.dat")
8| prec12 = sum_by_month(d12, months)
9| prec13 = sum_by_month(d13, months)
10| prec14 = sum_by_month(d14, months)
11|
12| create_bargraph("out.png", months,
13|     ["2012", "2013", "2014"],
14|     prec12, prec13, prec14)

```



\$ now run -e Tracker experiment.py



Project



experiment.py



precipitation.py



p12.dat



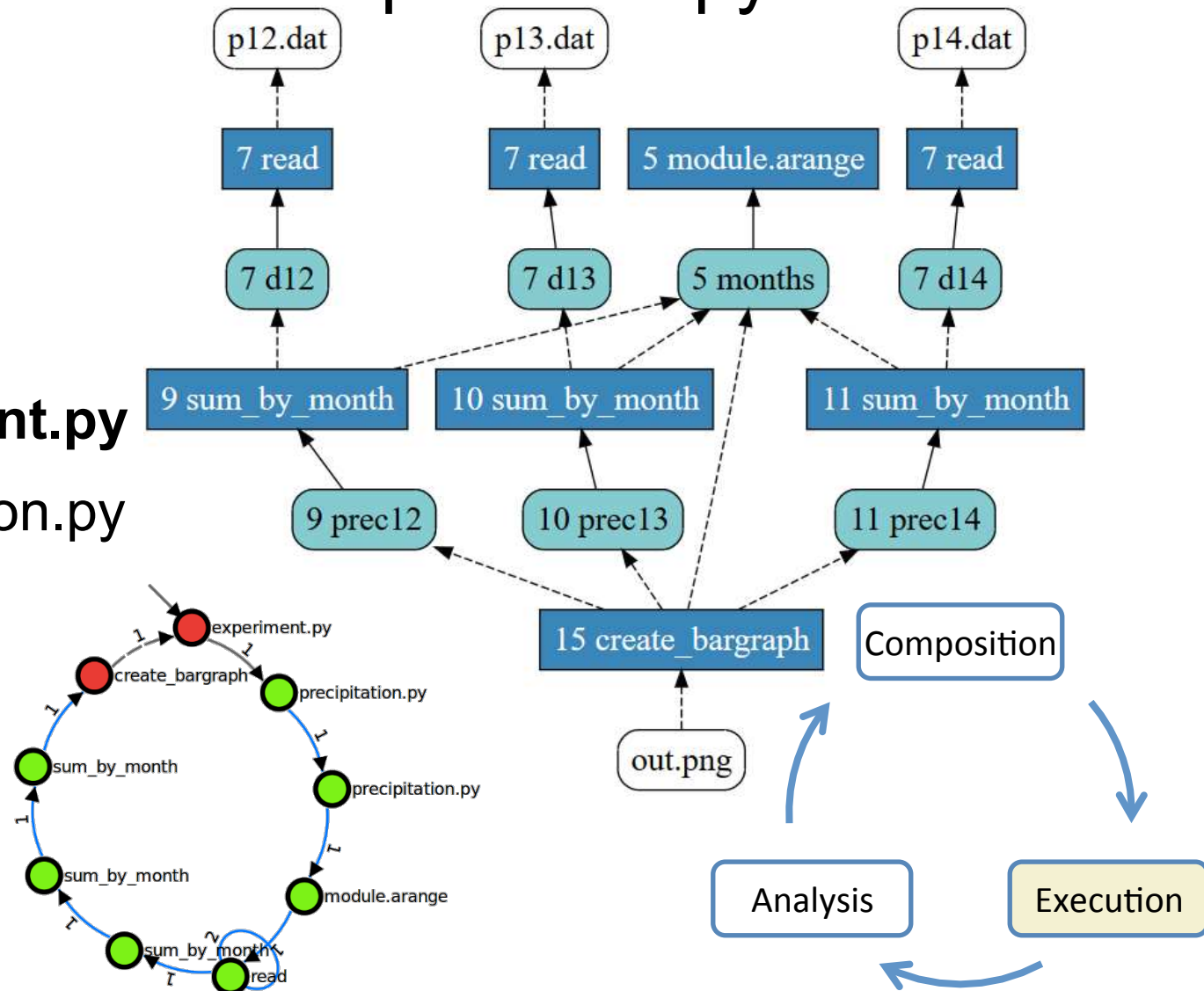
p13.dat



p14.dat

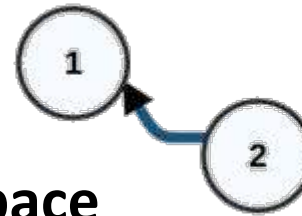


out.png




Version Model


Trial History





Product Space


 Project


 **experiment.py**

 precipitation.py

 p12.dat

 p13.dat

 p14.dat

 out.png

provenance

Version Space

Trial 1 Trial 2

1

2

1

1

1

1

1

1


1


1

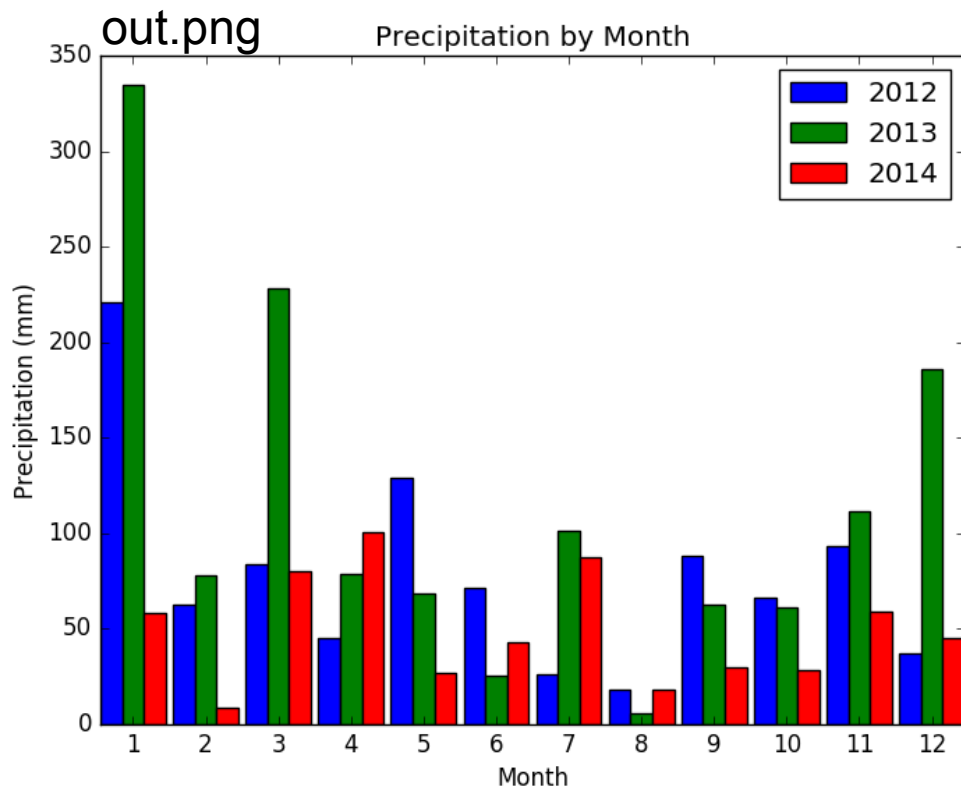
2

1

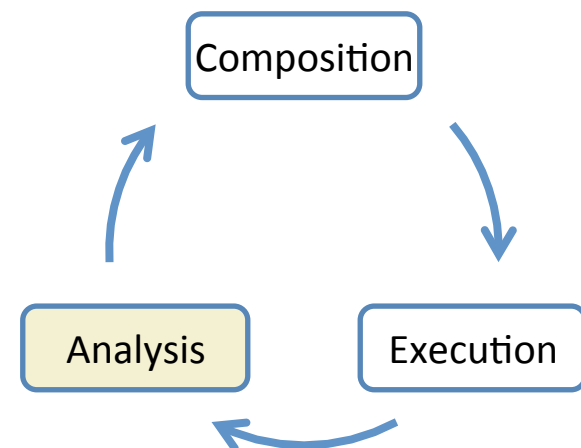
2

 write

 read



Conclusion:
“2012 was similar to 2013”

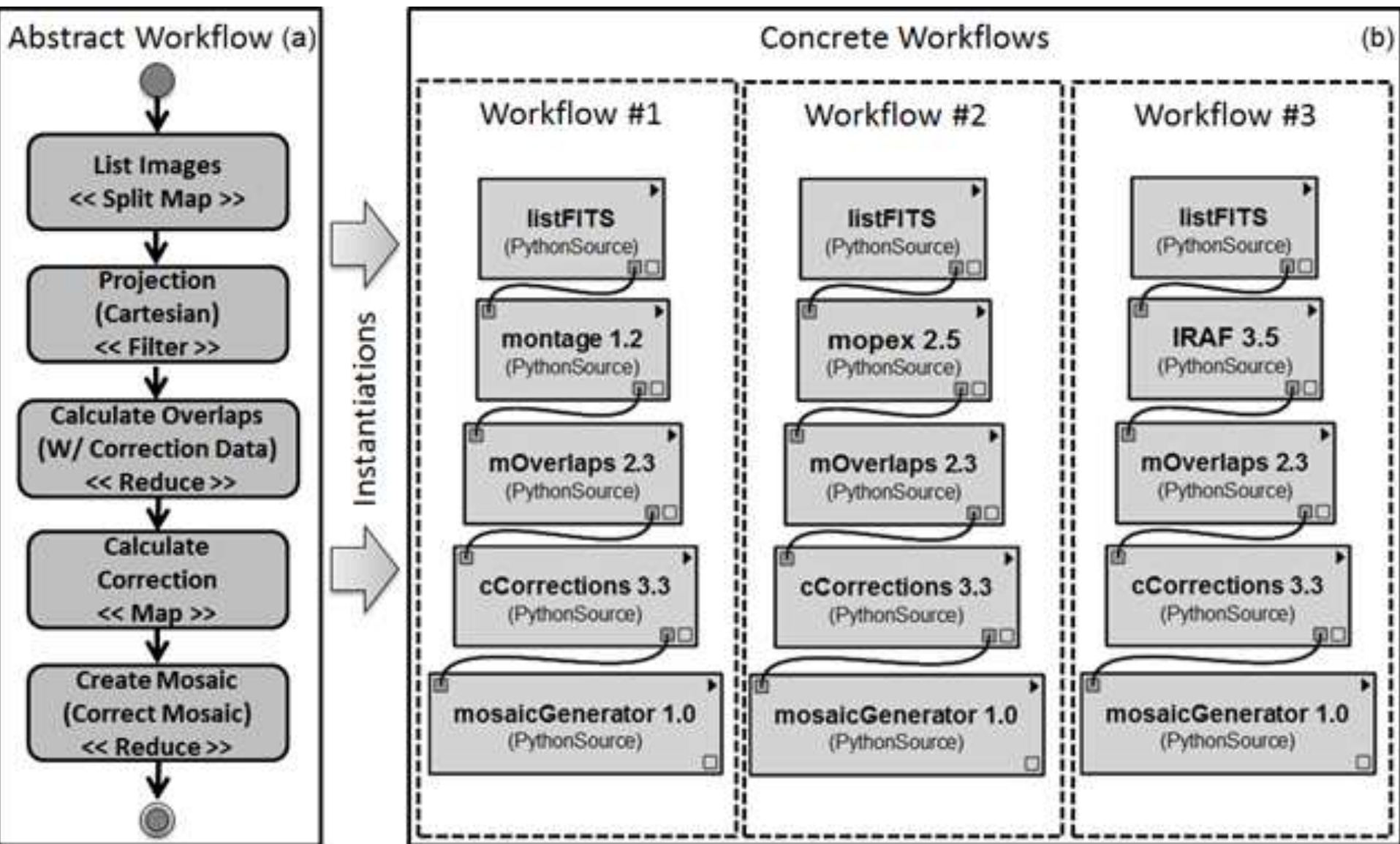


I don't think its enough to compare
just these years. Could you add
data from 2015?

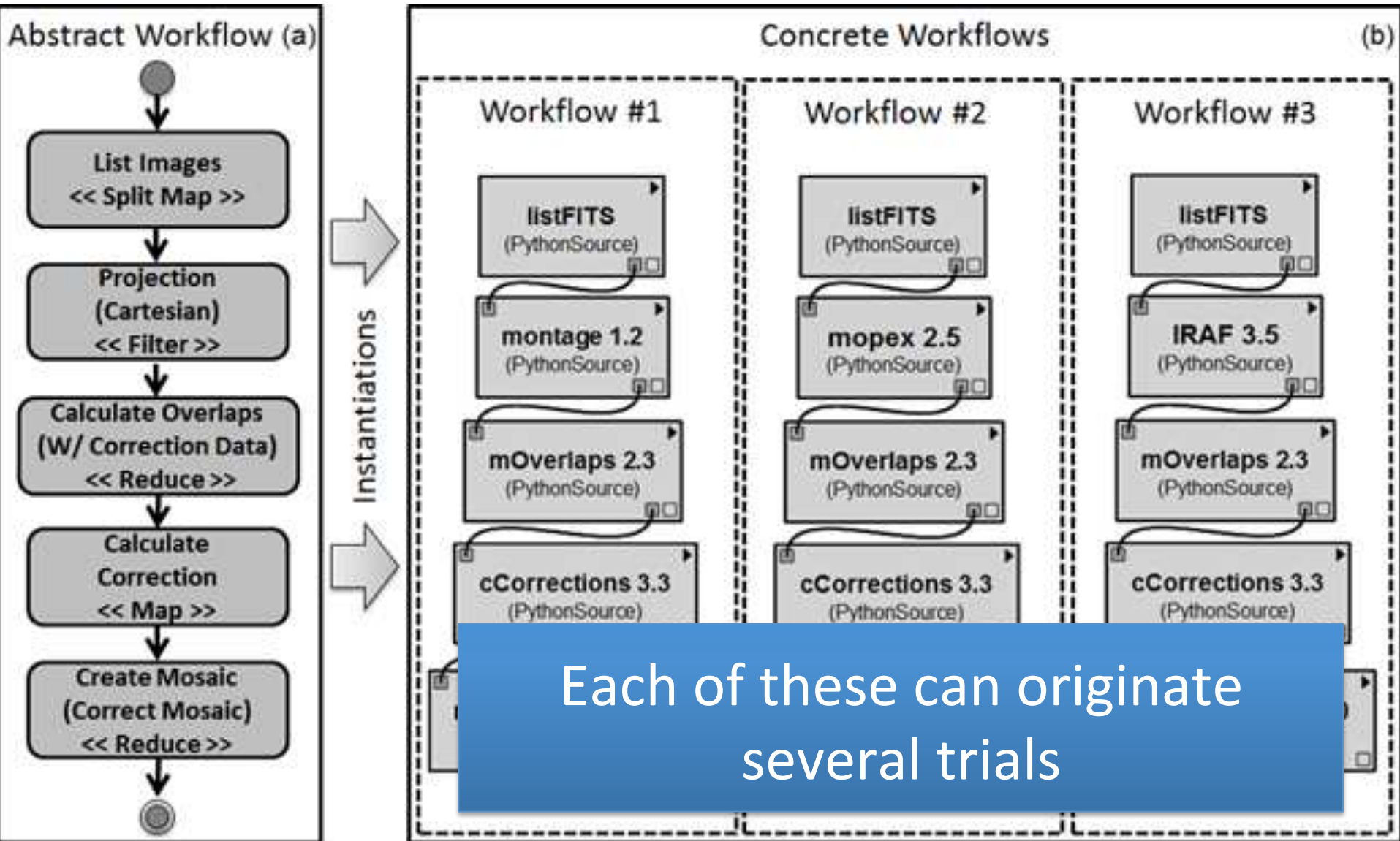


...

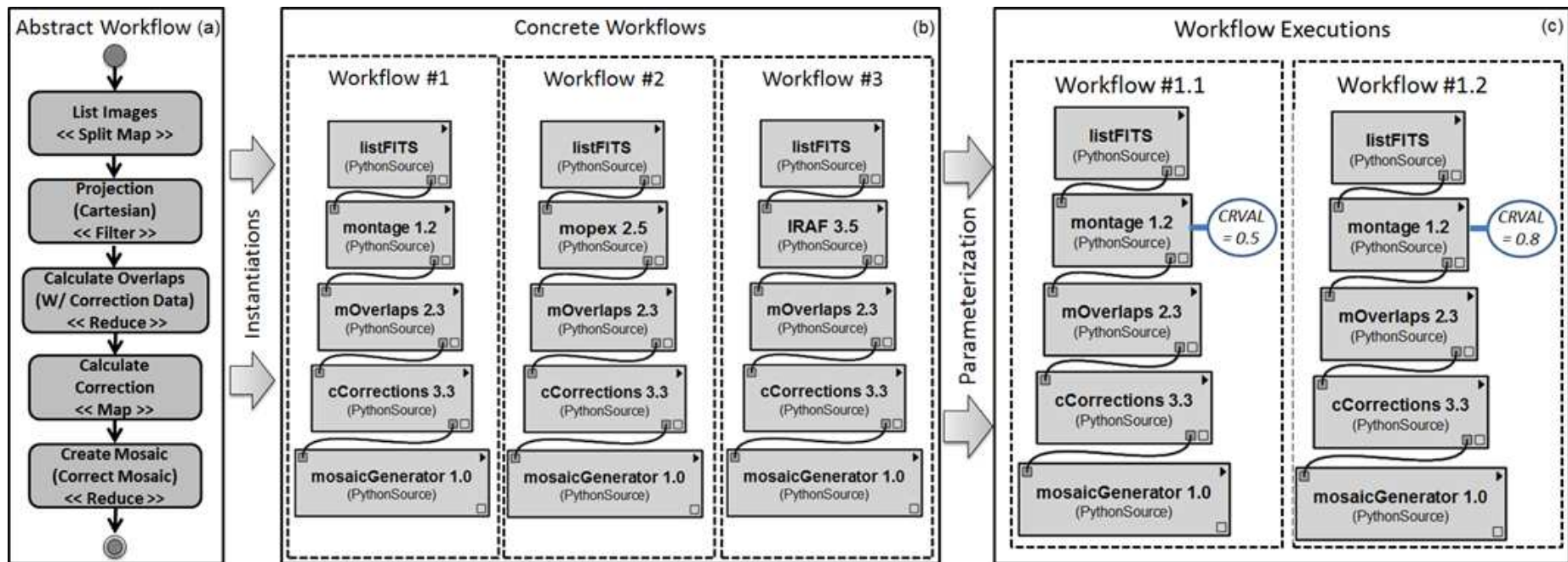
The same can be done for workflows



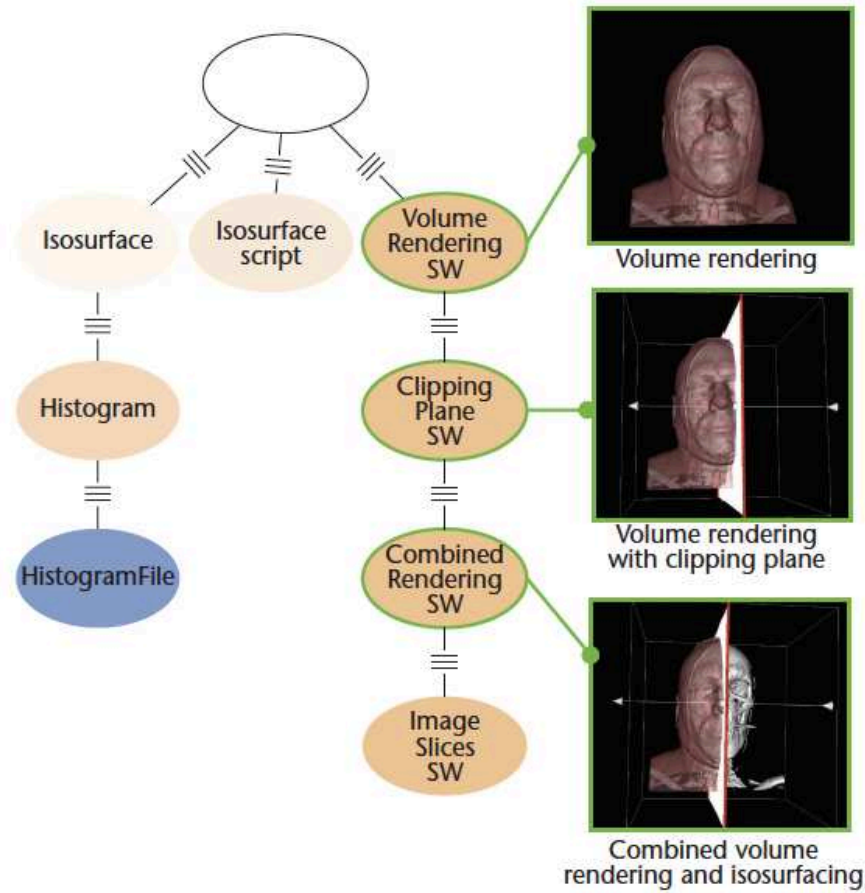
The same can be done for workflows



Trials in Workflows



History Graph (VisTrails)



Source: Freire et al., 2008. Provenance for Computational Tasks: A Survey.

Several ways to go from abstract to concrete

- When using scripts, there are several ways to go from abstract to concrete workflows
 - Activities are implemented one after the other in the script (no functions)
 - Activities are mapped into functions (each activity becomes one or more function)

Black Box X White Box

- In Workflow systems, activities are black boxes
 - What goes in and out are known, but what happens inside is not known
- In scripts, activities can be black boxes or white boxes
 - An activity in a script can call an external program, and in this the activity is a black box
 - When the function is implemented in Python, it is a white box

Black Box X White Box

- Black boxes have implications in provenance analysis

```
1| DRY_RUN = ...
2|
3| def process(number):
4|     while number >= 10:
5|         new_number, str_number = 0, str(number)
6|         for char in str_number:
7|             new_number += int(char) ** 2
8|         number = new_number
9|     return number
10|
11| def show(number):
12|     if number not in (1, 7):
13|         return "unhappy number"
14|     return "happy number"
15|
16| n = 2 ** 4000
17| final = process(n)
18| if DRY_RUN:
19|     final = 7
20| print(show(final))
```

Which values
influence the
result printed by
this script?
(variable **final**)

```

1| DRY_RUN = ...
2|
3| def process(number):
4|     while number >= 10:
5|         new_number, str_number = 0, str(number)
6|         for char in str_number:
7|             new_number += int(char) ** 2
8|         number = new_number
9|     return number
10|
11| def show(number):
12|     if number not in (1, 7):
13|         return "unhappy number"
14|     return "happy number"
15|
16| n = 2 ** 4000
17| final = process(n)
18| if DRY_RUN:
19|     final = 7
20| print(show(final))

```

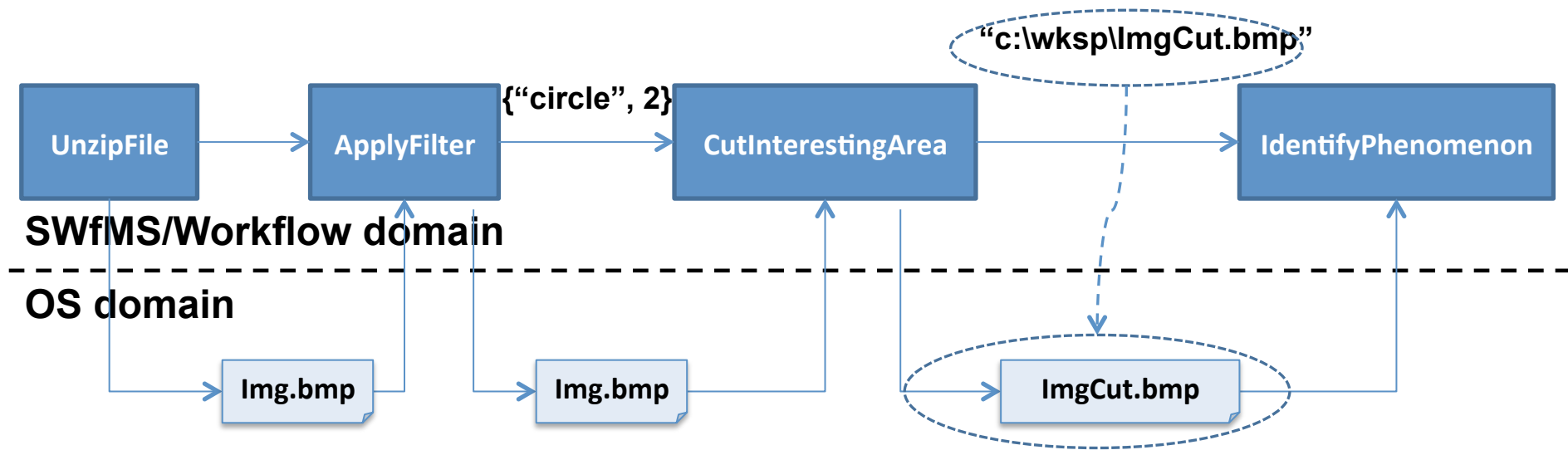
If DRY-RUN is 7, then
final depends only on
DRY_RUN

If not, then **final** also
depends on **n**

Implications of Black Boxes

- If **process(number)** were a black box, anything could happen inside it
- It could, for example, read a file that could influence the value returned by the function, so dependencies would be missed
- This is a common case of **implicit provenance**, that is missed by several provenance capturing approaches

Implicit Provenance



Sources:

Neves et al., 2017. Managing Provenance of Implicit Data Flows in Scientific Experiments.

Marinho et al., 2011. Challenges in managing implicit and abstract provenance data: experiences with ProvManager.

Implicit Provenance

- OS-Based approaches are able to capture this kind of provenance
- Other approaches need special components to handle it (e.g. PROVMONITOR)

Overview of Existing Systems

- Workflow Management Systems
- Provenance Management Systems for Scripts

Workflow Management Systems

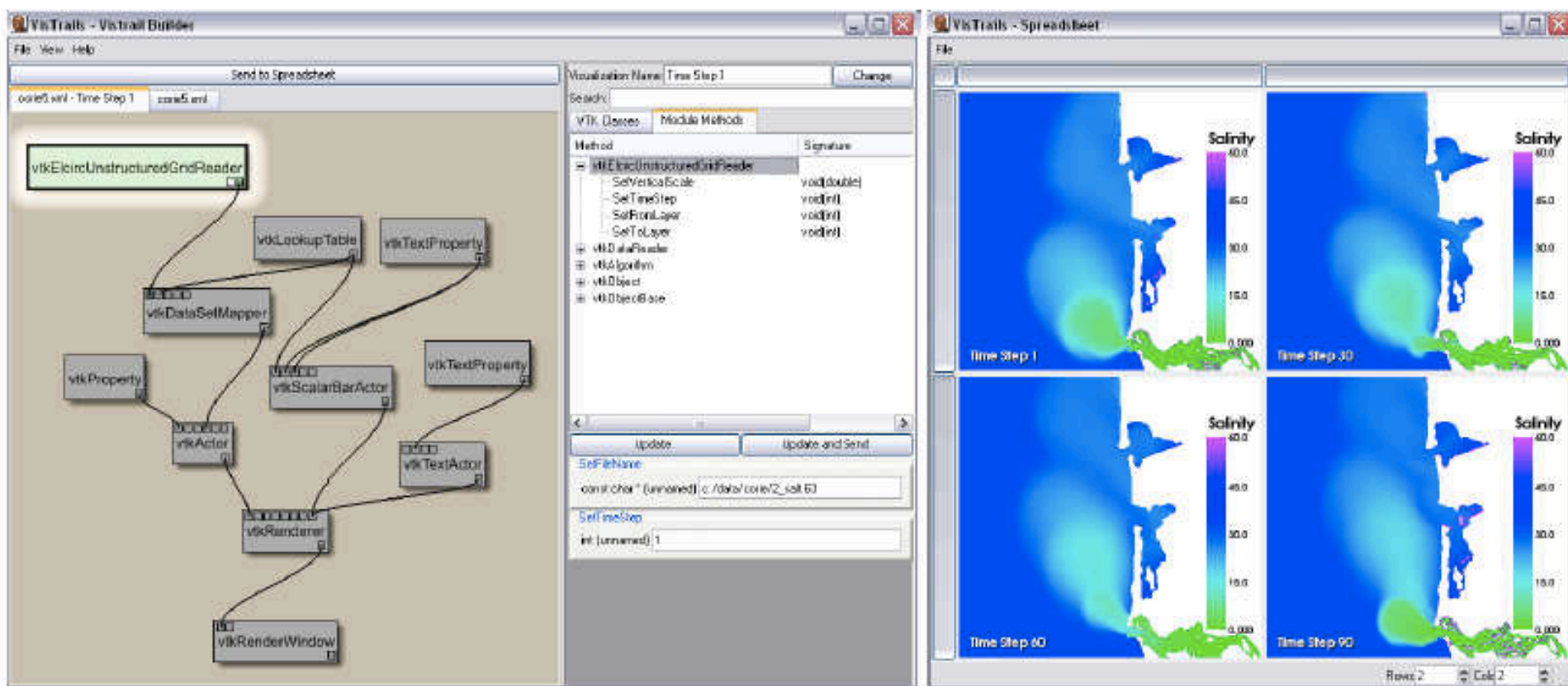
- VisTrails
- Taverna
- Kepler
- Swift
- SciCumulus
- Pegasus
- ...

VisTrails

- Visual drag and drop interface for workflow composition
- Captures history of changes in the workflow structure
- Allows comparing results side-by-side
- Focus on visualization



VisTrails

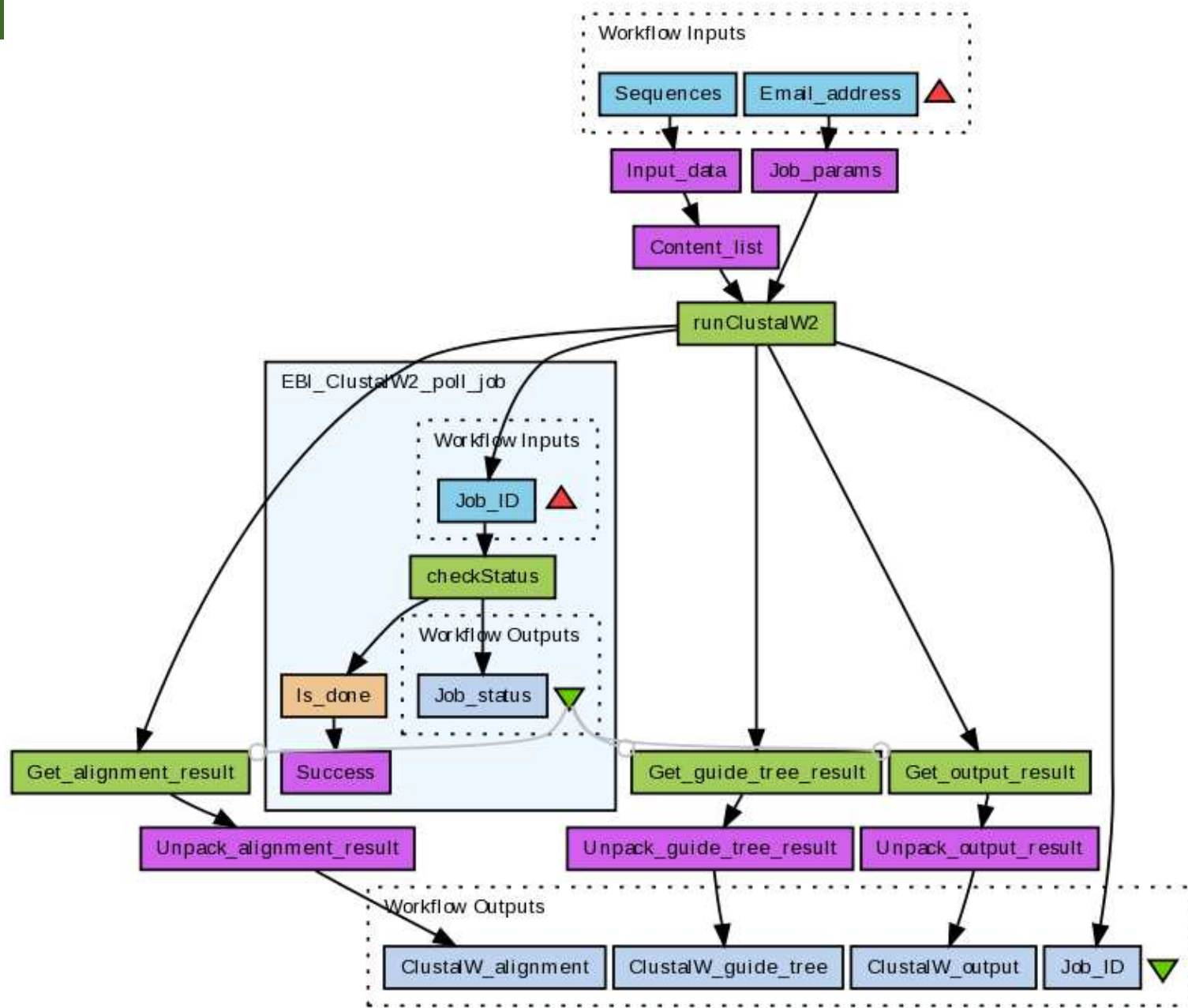


Taverna

- Focus on Bioinformatics
- Several ready-to-use bioinformatics services
- Drag and Drop graphical interface for workflow composition



<http://www.taverna.org.uk/>

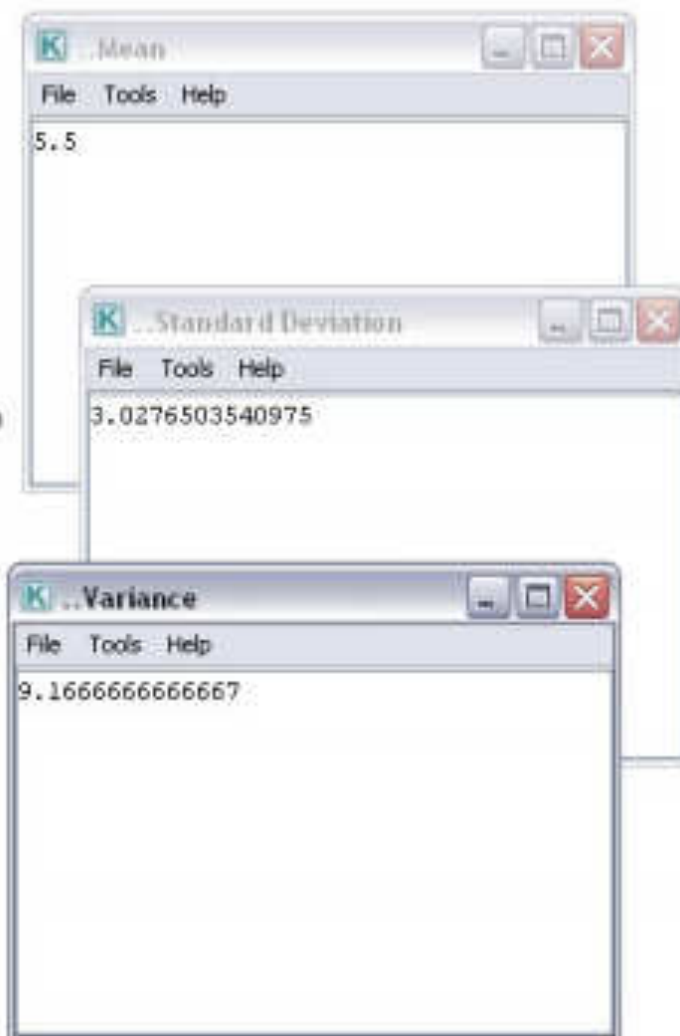
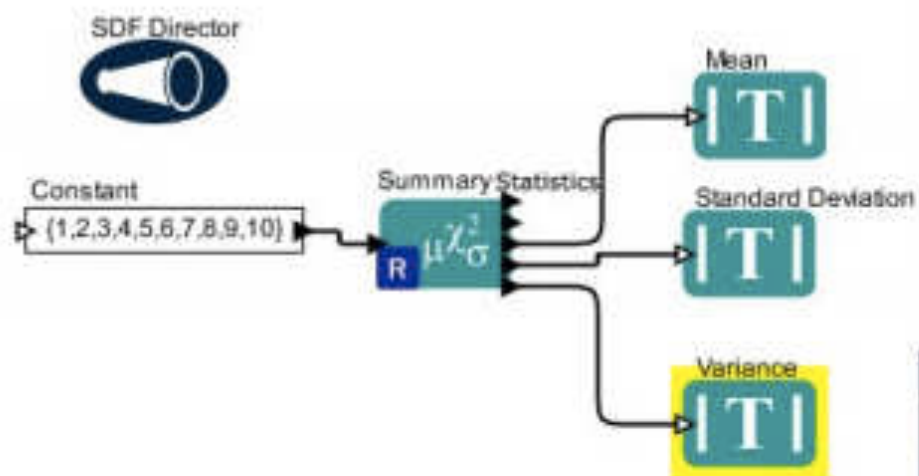


Kepler

- Drag and Drop graphical interface for workflow composition
- Different actors that rules how the workflow executed – Kepler workflows are not DAG



<https://kepler-project.org/>



Swift, SciCumulus and Pegasus

- Focus on High Performance
- Workflows are specified in XML (no graphical interface) in SciCumulus and Pegasus
- In Swift, workflows are specified as scripts in a specific language

<http://swift-lang.org/main/index.php>
<https://scicumulusc2.wordpress.com/>
<https://pegasus.isi.edu/>

Provenance Management Systems for Scripts

- noWorkflow
 - captures provenance for Python scripts
- RDataTracker
 - captures provenance for R scripts
- Sumatra
 - captures provenance for Python, R and MatLab scripts

Exercise

- Choose one of the systems presented in today's class and search the Web to find:
 - What is the format in which provenance is stored
 - If they export provenance in the PROV format

Provenance of these slides

- MARINHO, A. ; WERNER, C. M. L. ; MATTOSO, M. L. Q. ; BRAGANHOLO, V. ; MURTA, L. G. P. . Challenges in managing implicit and abstract provenance data: experiences with ProvManager. In: USENIX Workshop on the Theory and Practice of Provenance (TaPP), 2011, Heraklion, Creta, Grécia, p. 1-6.
- MATTOSO, M. L. Q. ; WERNER, C. M. L. ; TRAVASSOS, G. H. ; BRAGANHOLO, V. ; MURTA, L. G. P. ; OGASAWARA, E. ; OLIVEIRA, D. ; CRUZ, S. ; MARTINHO, W. . Towards Supporting the Life Cycle of Large Scale Scientific Experiments. International Journal of Business Process Integration and Management (Print), v. 5, p. 79-92, 2010.
- NEVES, V. C. ; OLIVEIRA, D. ; OCANA, K. A. ; BRAGANHOLO, V. ; MURTA, L. G. P. . Managing Provenance of Implicit Data Flows in Scientific Experiments. ACM Transactions on Internet Technology, 2017.
- PIMENTEL, J. F. N. ; FREIRE, J. ; BRAGANHOLO, V. ; MURTA, L. G. P. . Tracking and Analyzing the Evolution of Provenance from Scripts. In: International Provenance and Annotation Workshop (IPAW), 2016, Washington, D.C., v. 9672. p. 16-28.
- PIMENTEL, J. F. N. ; FREIRE, J. ; MURTA, L. G. P. ; BRAGANHOLO, V. . Fine-grained Provenance Collection over Scripts Through Program Slicing. In: International Provenance and Annotation Workshop (IPAW), 2016, Washington D.C., v. 9672. p. 199-203.