

OpenDreamKit

Work Package 4: User Interfaces



Overview

The general goals of WP4:

Improve existing Jupyter tools (T4.2)

Build new tools in the Jupyter ecosystem (T4.3, 4.8)

Improve OpenDreamKit components, especially where they interact with each other (T4.1, 4.4, 4.5, 4.12)

Demonstrate effectiveness in scientific applications (T4.9, 4.11, 4.13)

Work on Active Documents (T4.6, 4.7)



Overview

Goal: Provide unified interfaces for OpenDreamKit VREs

Broad categories of work:

- **Connect** with underlying computational software
- Notebook interfaces
 - Bring notebooks to **more communities**
 - Improve working with notebooks
 - **Collaborative** workspaces
- **Interactive** documents and documentation



Reporting Period 3

- **No new deliverables**, but software that people use is never complete!
- Instead, focus on supporting, improving work delivered earlier

Milestone M7: Prototype VRE for mathematical researchers

User story: The prototype VRE shall be extended with improved ease of deployment, new functionality such as interactive **3D visualization** and real-time **collaboration**, enabling researchers to collaborate productively in a **shared computational environment**. Finally, integrating notebooks and semantic knowledge into a publication / knowledge system enable a continuous process of leveraging OpenDreamKit components **from research to publication**.

OpenDreamKit produces tools for researchers to use. **WP4 is about making those tools accessible to researchers and research results accessible to others.**

OpenDreamKit bet on Jupyter notebooks

It has paid off!

- **Millions** of notebooks online (over 5M on GitHub alone)
- June 2018, Jupyter awarded prestigious 2017 **ACM Software System Award**
- Previous winners include: UNIX, TCP/IP, the Web, TeX, Java, GCC, LLVM

Background: Jupyter notebooks

Document with code, prose, maths, visualisation



Investigating the Sampling Theorem

In this section, we investigate the implications of the sampling theorem. Here is the usual statement of the theorem from wikipedia:

"If a function $x(t)$ contains no frequencies higher than B hertz, it is completely determined by giving its ordinates at a series of points spaced $1/(2B)$ seconds apart."

Since a function $x(t)$ is a function from the real line to the real line, there are uncountably many points between any two ordinates, so sampling is a massive reduction of data since it only takes a tiny number of points to completely characterize the function. This is a powerful idea worth exploring. In fact, we have seen this idea of reducing a function to a discrete set of numbers before in Fourier series expansions where (for periodic $x(t)$)

$$a_n = \frac{1}{T} \int_0^T x(t) \exp(-j\omega_n t) dt$$

with corresponding reconstruction as:

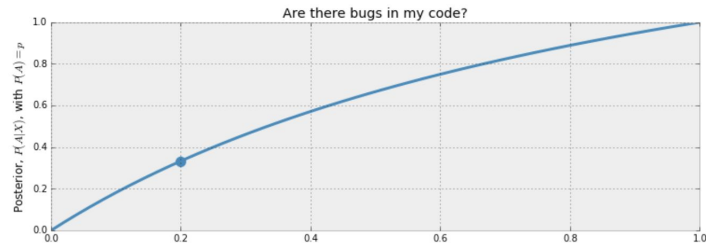
$$x(t) = \sum_n a_n \exp(j\omega_n t)$$

But here we are generating discrete points a_n by integrating over the **entire** function $x(t)$, not just evaluating it at a single point. This means we are collecting information about the entire function to compute a single discrete point a_n , whereas with sampling we are just taking individual points in isolation.

$$= \frac{2p}{1+p} \quad (11)$$

This is the posterior probability. What does it look like as a function of our prior, $p \in [0, 1]$?

```
figsize(12.5, 4)
p = np.linspace(0, 1, 50)
plt.plot(p, 2*p/(1+p), color="#348ABD", lw=3)
plt.title("Are there bugs in my code?");
```



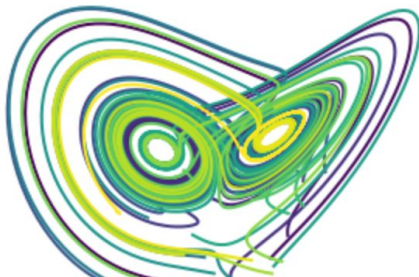
Background: Jupyter notebooks

It also has support for rich, **interactive** UI components:

```
: from notebooks.lorencz import solve_lorenz  
w=interactive(solve_lorenz,sigma=(0.0,50.0),rho=(0.0,50.0))  
w
```

sigma 11.50
beta 2.33
rho 21.80

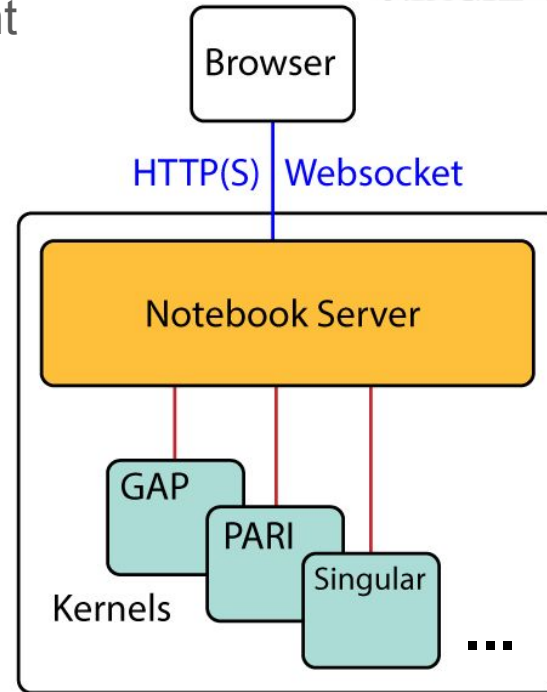
```
In [ ]: @interact  
def factor_xn(n=5):  
    display(Eq(x**n-1, factor(x**n-1)))
```



Background: Jupyter notebooks

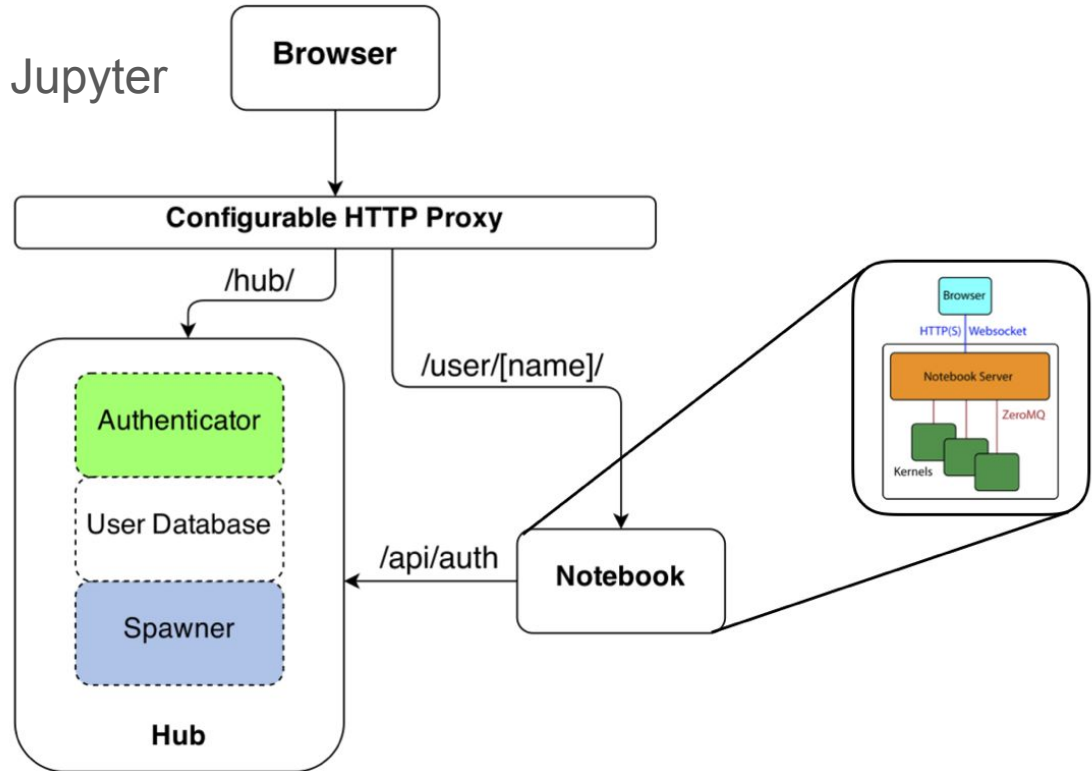
Web-based interactive computing environment

Language-agnostic protocol for computation



Background: JupyterHub

Extensible **VRE** built around Jupyter



Notebook Interfaces

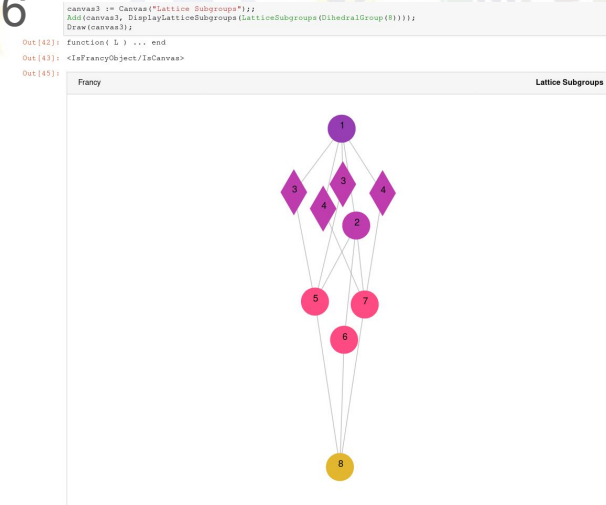
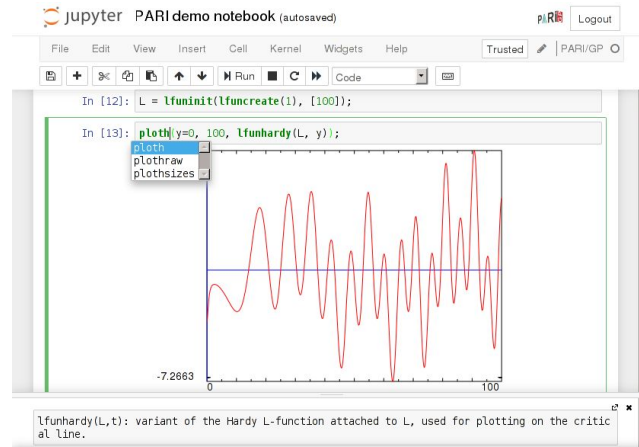
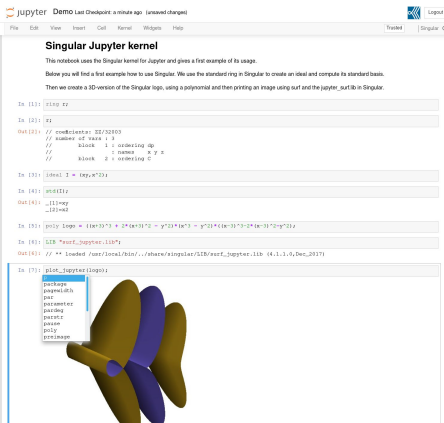
Key areas:

- Improve working *in* notebooks
 - Who can use notebooks?
 - What can be done in a notebook?
- Improve working *with* notebooks
 - What is difficult or unpleasant or impossible to do with a notebook?



Highlight: OpenDreamKit kernels

- Now **132 Jupyter kernels**, 8 contributed to by ODK.
- Further improved kernels from first reporting periods (GAP, PARI, Singular).
Delivered as D4.7, and MMT, GF, GLF as part of WP6
- Also contributed to kernels: cling (C++), SageMath



KPI: ODK Kernels on GitHub

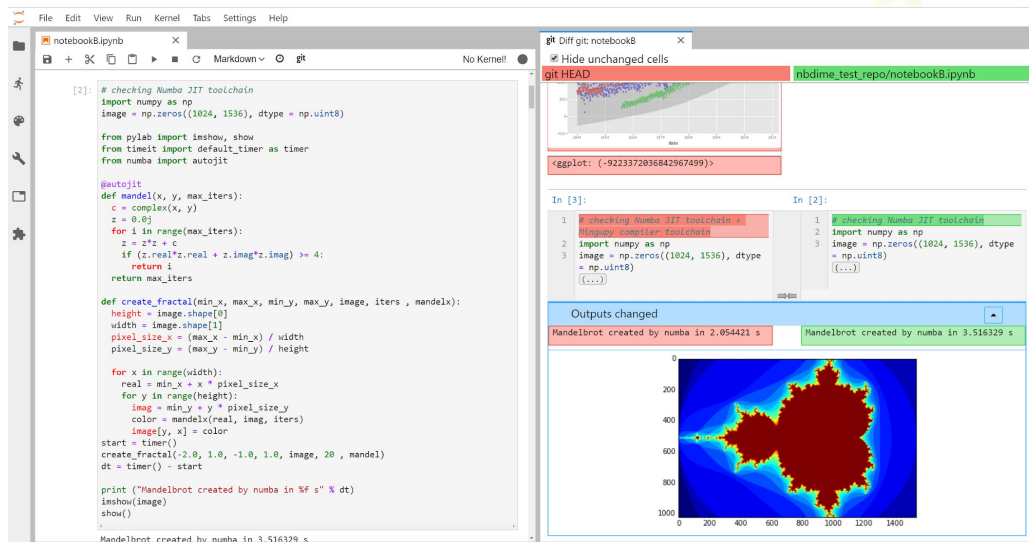
Notebooks found on GitHub using each kernel ([code](#)):

- SageMath: 8047
- Xeus-cling: 1216
- GAP: 94
- Singular: 11
- PARI/GP: 5
- MMT, GF, GLF: 5

Year	Total (incl non-ODK)	SageMath	Xeus-cling C++	GAP	Singular	PARI/GP	MMT
2016	467836		0	0	0	0	0
2017	1220829						0
2018	3087257	6199	684	63	8	3	1
2019	5540456	8047	1216	94	11	5	2

Highlight: nbdime

- Solves **major collaboration challenge** of change review (D4.6). Has been met with enthusiasm, adoption in the community.
- Jupyter Notebook and Jupyter Lab extensions, git integration.



The screenshot displays a Jupyter Notebook window with the following content:

```
[2]: # checking Numba JIT toolchain
import numpy as np
image = np.zeros((1024, 1536), dtype = np.uint8)

from pylab import imshow, show
from timeit import default_timer as timer
from numba import autojit

@autojit
def mandel(x, y, max_iters):
    c = complex(x, y)
    z = 0.0j
    for i in range(max_iters):
        z = z**2 + c
        if (z.real**2 + z.imag**2) >= 4:
            return i
    return max_iters

def create_fractal(min_x, max_x, min_y, max_y, image, iters, mandel):
    height = image.shape[0]
    width = image.shape[1]
    pixel_size_x = (max_x - min_x) / width
    pixel_size_y = (max_y - min_y) / height

    for x in range(width):
        real = min_x + x * pixel_size_x
        for y in range(height):
            imag = min_y + y * pixel_size_y
            color = mandel(real, imag, iters)
            image[y, x] = color
    start = timer()
    create_fractal(-2.0, 1.0, -1.0, 1.0, image, 20, mandel)
    dt = timer() - start

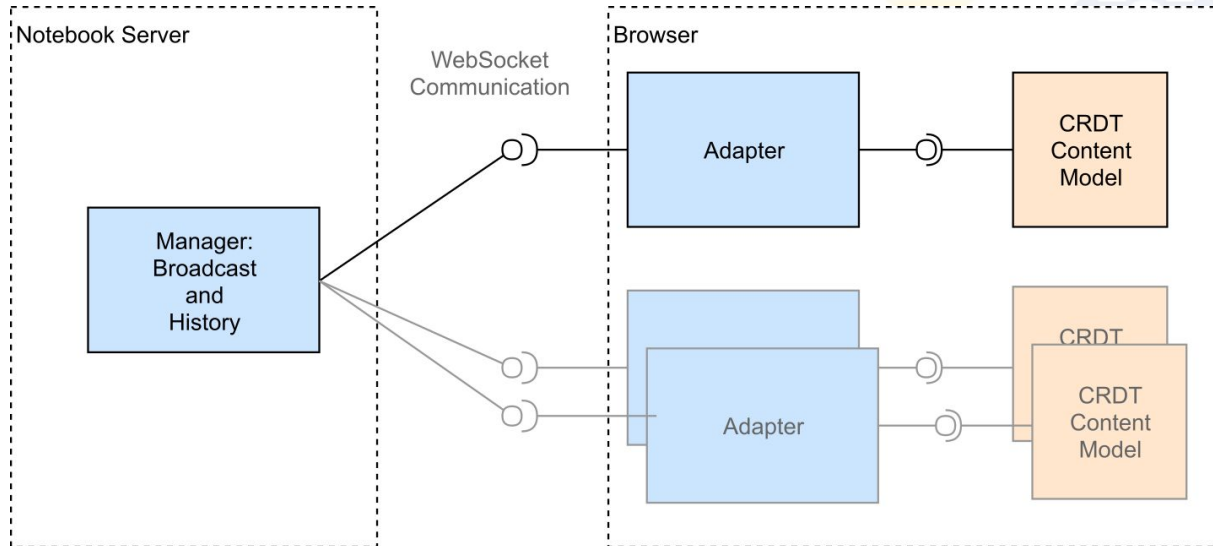
print ("Mandelbrot created by numba in %f s" % dt)
imshow(image)
show()

Mandelbrot created by numba in 3.516329 s
```

The right-hand side of the notebook shows the execution output, including a git diff window for 'nbdime_test_repo/notebook8.ipynb' and a plot of the Mandelbrot fractal. The plot is a 2D heatmap with axes ranging from 0 to 1400 on the x-axis and 0 to 1000 on the y-axis. The fractal is rendered in shades of red and yellow against a blue background. Below the plot, the execution time is displayed as 'Mandelbrot created by numba in 2.054421 s' and 'Mandelbrot created by numba in 3.516329 s'.

Highlight: real-time collaboration

- D4.15 prototype for live collaboration accepted as branch of main JupyterLab repo, continuing development. Required **extensive collaboration and coordination** with the JupyterLab team.



Highlight: real-time collaboration

Compressive_Sampling.ipyn

http://users.ece.gatech.edu/justin/ssp2

L_2 vs. L_1 Optimization

The classic constrained least squares problem is the following:

$$\min \|\mathbf{x}\|_2^2$$

with corresponding solution illustrated below.

```
[8]: from __future__ import division
      from matplotlib.patches import Circle
      x1 = linspace(-1,1,10)
      x1
```

```
[9]: draw_example()
```

1 Python 3 | Mode: Ln 5, Compressive_Sampling.ip

on acc
quired e
am.

Compressive_Sampling.ipyn

L_2 vs. L_1 Optimization

The classic constrained least squares problem is the following:

$$\min \|\mathbf{x}\|_2^2$$

with corresponding solution illustrated below.

```
[8]: from __future__ import division
      from matplotlib.patches import Circle
      x1 = linspace(-1,1,10)
      x1
```

```
[9]: draw_example()
```

Note that the line is the constraint so that any solution to this problem must be on this line (i.e. satisfy the constraint). The L_1 solution is the one

0 Python 3 | Mode: Ln 4, Compressive_Sampling.ip

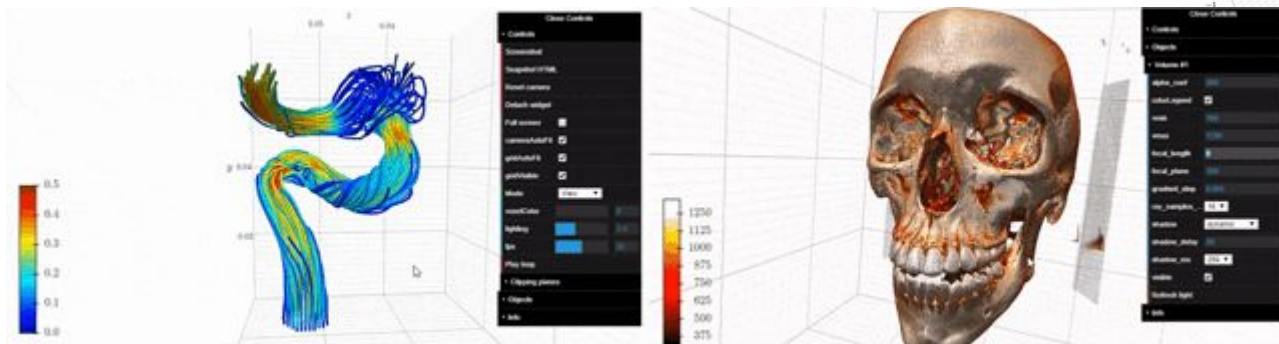
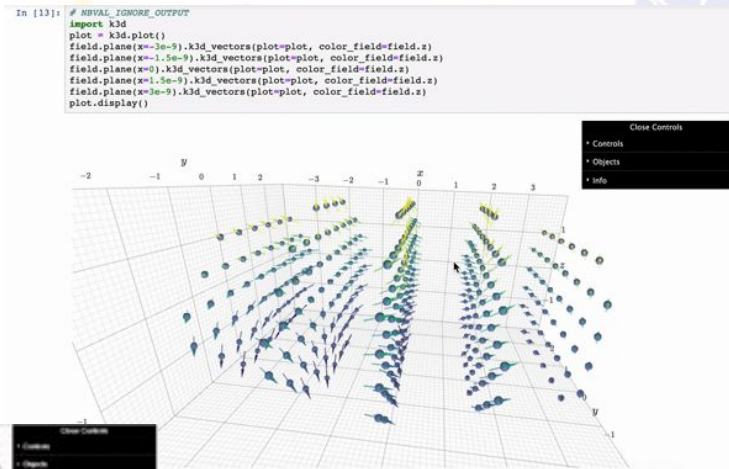
Browser

Adapter

Content Model

Highlight: 3D visualisation in Jupyter notebooks

- D4.12: Jupyter extension for 3D vis, demonstrated with fluid dynamics, micromagnetics
- Packages:
 - k3d-jupyter
 - ipydatawidgets
 - ipyscales
 - unray
- Solves major challenge for remote VREs



Highlight: Dynamic documentation and exploration system

- Delivered D4.16 (RP2) as two new packages, built on Jupyter widgets for interactively exploring objects in SageMath
 - Sage Combinat Widgets
 - Sage Explorer

```
1 # A Grid View for a Young tableau
2 from sage_combinat_widgets import GridViewWidget
3 t = StandardTableaux(15).random_element()
4 wt = GridViewWidget(t)
5 wt
```

1	3	5	6	7	8	11	12	-	I	+
2	10	13	-	+						
4	14	-	+							
9		+								
15	-									
										+

Interactive Documents

Key areas:

- Active Documents
- Interactive Documentation



Highlight: Active Documents Workshop

Workshop on live documents hosted in Oslo. Resulted in new package: **thebelab**, for embedding execution on any page, integrating tools from JupyterLab and MyBinder.org, and support for notebooks in **MathHub portal**, jupyter.mathhub.info

Linear algebra

```
A = matrix(GF(7), 4, [5,5,4,3,0,3,3,4,0,1,5,4,6,0,6,3]); A
```

run

```
P = A.characteristic_polynomial(); P
```

run

```
P(A)
```

run

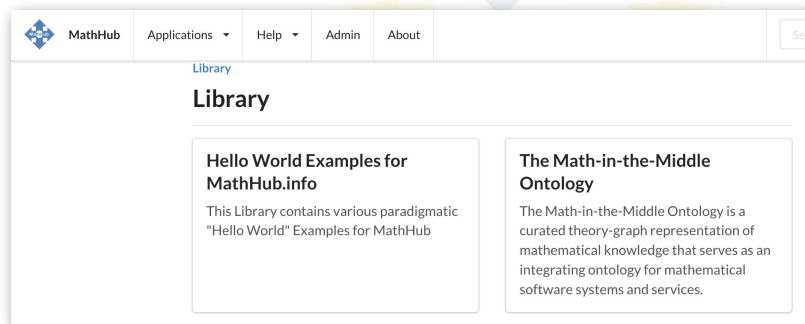
```
A.eigenspaces_left()
```

run

Computing the rank of a large sparse matrix:

```
M = random_matrix(GF(7), 10000, sparse=True, density=3/10000)
M.rank() # random
```

run



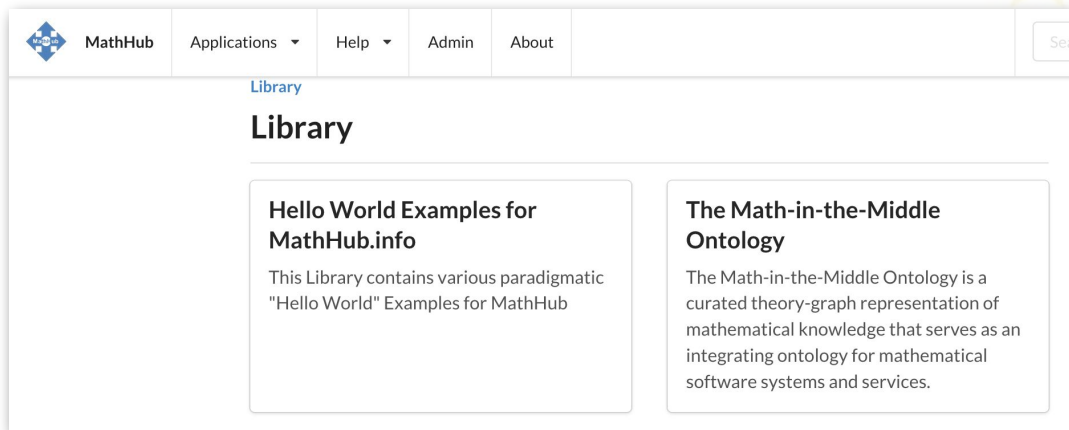
The screenshot shows the MathHub portal interface. At the top is a navigation bar with the MathHub logo, 'Applications', 'Help', 'Admin', and 'About' menus, and a search box. Below the navigation bar is a 'Library' section. The 'Library' section contains two featured items:

- Hello World Examples for MathHub.info**: This Library contains various paradigmatic "Hello World" Examples for MathHub.
- The Math-in-the-Middle Ontology**: The Math-in-the-Middle Ontology is a curated theory-graph representation of mathematical knowledge that serves as an integrating ontology for mathematical software systems and services.

Highlight: MathHub notebook integration

MathHub.info is a portal for active mathematical documents. As part of D4.11, a notebook integration with MathHub was added. This allows:

- Authoring MathHub documents as Notebooks
- **Interactively** exploring existing MathHub documents as a Notebook.



Highlight: Core infrastructure

- Refactorisation of SageMath's Sphinx **documentation** system as part of D4.13
- Improve Sphinx support for Cython projects.
 - Enabled building proper documentation for fpylll, CyPari2, CySignals.
- CyPari2 used to be part of SageMath, but it was made a separate package in D4.10 (see also D4.1). This ties into **WP3**.
- To completely enable Cython documentation out of the box, Python needs to be fixed. For this, we submitted PEP (Python Enhancement Proposal) 580. Benefits not only Sage, but **all Python users**.

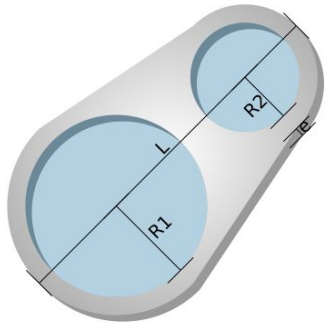
KPI: JupyterHub deployments

- Local CoCalc instance at **Universität Zürich**.
 - Deployed September 2015 - February 2016.
 - People involved: @pdehaye, @williamstein
- **Instance of JupyterHub** deployed by the **Mathrice group**
 - Host Infrastructure: France Grille's **LAL cloud**
 - Users: members of math labs in France
 - Main use case: casual use
- Local **JupyterHub instance at Université Paris Sud / Paris Saclay**
 - Host Infrastructure: France Grille's LAL cloud
 - Users: personnel and students of UPSud / Paris Saclay
 - Main use case: use in classroom (Python, Sage, C++), casual use
 - People involved: @VivianePons, @nthiery, @gouarin
- **JupyterHub instance deployed on USheffield's HPC system**
 - People involved: @mikecroucher
- **JupyterHub instance(s) deployed at UVSQ**
 - Main use case: use in classroom (Sage, Python, C, Apache Spark), casual use
 - People involved: @defeo
- JupyterHub and Binder instances deployed on **EGI** infrastructure, as well as **OVH, Azure, Google, and GESIS**
- Easy deployment of live GAP/SageMath/... notebooks with **Binder**, thanks to the Docker containers (#58); potential alternatives: **Debian packaging** and **Conda packaging**.
 - People involved: @nthiery, @minrk, ...
- Local instance of CoCalc (using the Docker container) at the **University of Gent**
 - Main use case: teaching for mathematics students
- Deployed at **jupyter.mathhub.info**
 - With MMT kernel
 - People involved: @tkw1536

Highlight: Simulagora

- Logilab **VRE** deployment for application development and deployment.
- Can use **JupyterLab** for application development, which can then be deployed with a simplified parameters form input.

Outils de calcul mécanique



Length (L) mm

Radius 1 (R1) mm

Radius 2 (R2) mm

Thickness (e) mm

Geometry Measures Summary

Geometry: RRA CPY

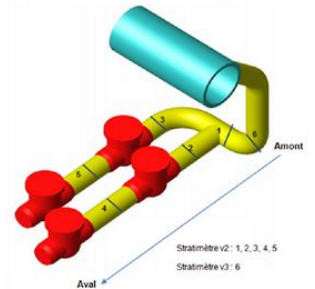
Material name: Z2CND1712

User material

Material name:

The material must be defined in the user data directory.

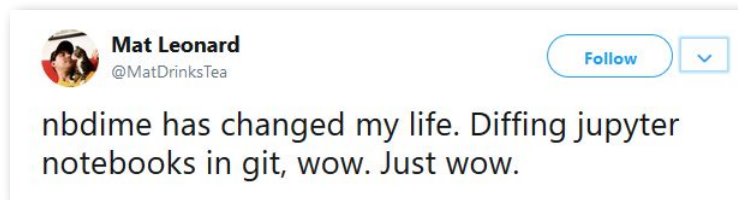
Add to the user list



Submit

KPI: Usage/impact statistics (since last reporting period)

- nb dime: 1336 stars on github (855 at RP2), 61 contributors (45 in 12 months prior), 388 comments, 107 new issues (94 closed).
- ThebeLab: 96 stars (44 at RP2), 28 contributors (15 in prev 12 months), 124 new issues (12 closed), 292 comments.
- K3D-Jupyter: 154 stars (48 at RP2), 14 contributors, 140 new issues (129 closed), 303 comments.



Summary

- ODK has contributed a **VRE toolkit**
- Jupyter provides the **interface** for these VREs
- Thanks to ODK,
 - **More communities** can use Jupyter
 - SageMath and other ODK software can be used in **more contexts**
 - **Collaboration** in Jupyter is greatly improved
- We have a **demonstration VRE**, built with this toolkit, used for upcoming demos



end

Will present some high-level lessons and future ideas after the demos



Work Package 4

Lessons and Future



Jupyter and the Future

- How did Jupyter contribute to the success of ODK
- Funding determines the direction of project
- Stakeholders are solving their own problems (Banks, large Enterprise)
- Core development still needs support
- EU can ensure that Jupyter core continues to serve Europe, Academia

Jupyter and ODK

- Building on Jupyter has enabled **sharing** and **dissemination** of mathematics research in ODK
- New mathematics communities have **access** to interactive computing and visualisation that didn't before
- **Education** and **research** facilitated by improved interactive capabilities
- Several **major challenges** of coordinating with open source projects with many stakeholders, but **big rewards** in the end, especially for long-term **sustainability**
- Web-based Jupyter makes it easier to build **cloud-based VREs** with ODK

Before OpenDreamKit

- Jupyter showed promise as a **shareable** document and **interactive** environment
- Only available to certain programming communities (Python, Sage, Julia, R, etc.)
- However, there were **key gaps** in the collaboration workflow
- **Cloud/remote** computing growing, but desktop capabilities were lost
- Many opportunities for interactive **explorations**

After OpenDreamKit

- Jupyter has proven valuable for sharing and interactive computing
- Now available to many **more communities** (ODK kernels)
- Gaps in collaborative workflow have been filled (**nbdime, nbval, JupyterLab**)
- **3d visualization** enabled in Jupyter via **K3D, ipydatawidgets, pythreejs**.
- Improved interactive **exploration** via widgets
- Improvements throughout the Jupyter and Python ecosystems along the way

Jupyter and the Future

- Building on **existing projects** has enabled ODK to have high impact
- **Contributors tend to solve their own problems.** Corporate participation in Jupyter is valuable, but priorities often differ from academia and public good
- **Core project development** still needs long-term funding to be secure
- Funding core development **enables EU to guide direction** of Jupyter or similar projects and **support STEM** research and education through key infrastructure
- ODK is special because it includes **explicit funding for core software infrastructure.** There should be more of this!

Binder: Jupyter for Open Science

- Submitted BOSSEE proposal to **EOSC** to fund Binder for Open Science (not funded)

Binder currently hosted by:

- Jupyter Team (US, Google Cloud)
- OVH (France, cloud provider)
- GESIS (Germany, social sciences institute)
- Turing Institute (UK, open science advocate)

Thousands of people are using Binder every day

**Binder is the frontier of Jupyter in
Open and Reproducible Science**

