



K HELIX

Hellenic Data Service





HELIX

Hellenic Data Service

Spiros Athanasiou

IMSI/Athena RC



Research & Innovation Information Technologies

Athena RC Why us?

- Our scientific focus is on cross-disciplinary, transformative data-intensive research (Big, Open, Linked data)
- growth
- (scientific infrastructures, data catalogues)
- Our collective insights and knowledge shaped the vision, implementation, an governance of HELIX grounding it on real-world challenges and considerations



• We champion **Data Economy**, **Big Data** and **Data Science** for national economic

We lead EU/national policies and technical interventions on Open Access/Data

Motivation

Motivation

Converging Policy landscape

- **Data Economy** a strategic priority for EU's sustainable future growth integrating policy, technology, and innovation actions
- **Public Sector Information** open up and create value from public-sector and publicly-funded Data (open data, INSPIRE, H2020, OGP, ...)
- **Industrial Data Platforms** emerging organization & technical instrument to facilitate data sharing and valorization within EU industrial value chains
- Research
 - **Open Access** de jure policy for sharing EU-funded scientific output
 - **Data Management Plans** formalize data handling on project/organization-level
 - **FAIR data** de facto international policy for scientific data



Motivation **The Big Picture**

Economic growth, scientific progress, and societal prosperity are about searching, sharing, using, experimenting, building, and valorizing



(*frictionless)

Data

* = simple, fast, inclusive,

Motivation

Archiving-focused Data Platforms

- Research Data Platforms heavily focused by the priorities and workflows of **Digital** Libraries (i.e., serve archiving and provenance)
 - Introducing artificial barriers for researchers
 - Narrowing real-world relevance
 - Most useful data are not linked with publications
 - Make it easier to publish data, why the strict rules?
 - Help me use and experiment with data



am not a **ibrarian**

Motivation Challenging the status quo

- We need flexible, low-cost, open, collaborative services for simplifying sharing, discovery, use, analysis, and visualization of scientific data
- Let's change: the explicit assumption is that RDPs must serve scientists



ama scientist

Motivation Why is this needed?

- Research Data Platforms

 - lifecycle
 - and services
 - Ensure **sustainability** •



Key lessons learned from Open Data are highly relevant for

• Lower the entry barrier, making it easy, simple, and fast to publish and find data • No walled gardens; all data, from any field are welcomed, at any point of their

Make data useful to more people, most of the time (80/20) through visualization

Motivation **Sustainability**

- needs of scientists

 - Relatively low CAPEX (setup), higher and fluctuating OPEX (operation, growth) • Public funding may not **suffice** or be timely **available**
 - **Devaluation** is (only) a few steps away (stale/missing data, no QA/SLA)
 - Need to introduce additional **revenue streams**, but from where?
- Industry-relevance (another lesson from Open Data)
 - Industry amongst the first and leading users of Open Data, generating value from new/improved services



• A Research Data Platform must be **diachronic**, ensuring data are always accessible, and evolving, addressing the ever-growing data-intensive

• Sharing and using industrial data in commercial value chains remains a challenge

Motivation **Industrial Data Platforms**

- ensuring fair reimbursement of industrial data
 - cost)
 - IPR protection)
- We can inherently serve these needs, provide a parallel industrial data platform by-design, and tap into the additional revenue streams
 - & open data, data science as-a-service (DataSaaS)



Data Platforms for securely sharing, discovering, licensing, using, and

• Concept follows the paradigm of open data (simplicity, fit for purpose, benefits, fast, low

• Same technical **foundations** with key differences (confidentiality, contract management,

USPs: scalable production-grade data processing/analysis services, unified proprietary



Service

HELIX **Hellenic Data Service**

- Scientific elnfrastructure for data-intensive research
 - Supports the full lifecycle of scientific data management, processing, sharing, and reuse
 - Inherently scalable, cloud-based
 - Nation-wide, horizontal, cross-domain
 - Low-cost, economies of scale, network effects, maximize ROI
 - Multiple roles: Open Access, FAIR Data, Public Data, Industrial Data Platform



Data first

HELIX The 3 pillars of HELIX

Publications

- Nation-wide, cross-domain discovery of publications
- Adapt and localize **OA OpenAIRE CRIS** services
- Data
 - Data catalogue and repository for FAIR scientific and industrial data
 - Discover, collect, evaluate, download, and use

• Labs

Generic-purpose and domain-specific services and APIs for data analysis, processing, and experimentation



Data alone is not enough



HELIX **Target groups**

- Scientists: data sharing, OA publishing, data experimentation
 - All scientific fields, including **citizen scientists**
- **Organizations:** institution-wide services augmenting, exposing, or replacing existing publication & data catalogues/repositories
 - Academia, Research, Public Administrations (PSI), special-interest groups
- Scientific Infrastructures: building block; scalable data processing services for very large, heterogeneous scientific data
 - Upcoming: ELIXIR (bio), APOLLONIS (linguistic) •
- Industry & innovators: value-added services; ad hoc analysis services •
 - Industrial Data Platform: low-cost data processing infrastructures; Data Science as a Service, • training data for ML



HELIX Core Concepts 1/2

- **Data-first:** make it simple, easy, and fast to share data (<10 secs); this is what is truly missing; build critical mass (data & users)
- Scientists first: serve the scientists, not librarians or standardization bodies; all too often this is lost, raising the entry barrier and thus failing (see open data)
- Just another tool: ensure inclusiveness and downplay our potential impact on the scientific process be useful and in the background (just another hammer)
- Love ALL data: any data used during research (not only in pubs); we do not know what/how/where data will be useful; no data is too little, no data is too small



HELIX **Core Concepts** 2/2

- is fluid
- fast visualization);
- give back)
- design and development of the system itself
- integral in modern scientific practice



• Cross-disciplinary: actively avoid walled-gardens and domain silos; facilitate data-driven cross-disciplinary research (introduce data & problems, facilitate networking); 'scientist' role

Bundle data with services: software, tools, and knowhow on how to use data is the 2nd greatest bottleneck behind data availability; think equally big (e.g., Spark) and small (e.g.,

Openness as a principle: open software, open standards, open services (learn from others,

• Agility: flexibility and reusability across all provided services and sub-systems; also during

All Scientists are Data Scientists: data management, processing and analysis skills are

HELIX **Development Roadmap**

• Phase 0 (incubation): 2012-2017

Infrastructures Roadmap

• Phase 1 (MVP): 2018-2019

- Phase 2 (Beta): 2020-2024
 - clients; governance structure; industrial data platform
- Phase 3 (Production): 2025-
 - Sustainable diachronic operation



• Original concept & funding proposal; core technology developed in other R&D projects; National Research

• MVP for technical/policy foundations; core services & lighthouse apps/communities; prepare follow-up

• Scale services and expand reach to more scientific communities; integration in 3rd infrastructures; first industrial

HELIX Architecture **Data-drive & Agile development**







HELIX Architecture

HELIX Architecture Three pillars

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pubs.hellenicdataservice.gr

data.hellenicdataservice.gr



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lab.hellenicdataservice.gr

HELIX Architecture Birds-eye view





HELIX Architecture **Core Principles**

- components
 - development tracks
 - Independently scale as/when/where needed, no single-point of failure
 - Workflow orchestration, management & monitoring via in-house Spring Boot sub-system •
- Cloud-based
 - Leverage and valorize GRNET's laaS cloud (knossos-okeanos) & HPC (ARIS) •
 - Docker-based, ported to Kubernetes \bullet
- Open Source/Open Standards

 - scale systems
- Shibboleth-based federated authentication for members of the Greek scientific community
 - •



Not a single monolithic system, but an assembly of loosely coupled, highly-scalable independent

Repurpose/extend software/APIs, flexible prototyping & experimentation, asynchronous parallel

Exclusively open: build on existing great software, give back to the community, help others Majority of software originally developed in the context of EU/national R&D projects, powering world-

Authorization handled individually by each applications by custom roles (SSO not advisable)

HELIX Architecture **Logical Architecture**







HELIX Architecture The data lifecycle







HELIX Architecture **HELIX-Core**

Entry point for discovering all HELIX services, resources, and guides

- Provides the illusion of a single application (common theme)
- Direct entry points also available (e.g., data.helix.gr)
- Loose, API-based integration of search results for all other services (Pubs, Data, Lab)
- Custom Spring app
 - Workflow management (data ingestion, housekeeping)
 - WordPress (content management)
 - Services/code reused in other services for AAI, multilinguality support, monitoring/logging





Latest news



About Helix

Research for Data

Find, access, and reuse data from Australian research organisations, agencies and institutions via Hellix flagship service.

Find research data

24 Anp 2018

Helix data principles: how well known or understood are they?

Explore the issues around health data. Learning activities specifically for people working with medi... Read more

Our partners

Helix collaborates with universities and other research institutions to enhance the value of data and enable new discoveries.

Who we work with

About

Αρχική

Data Publications

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Το έργο Επικοινωνία Όροι χρήσης

Lab Θεματικές Οργανισμοί

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24 Amp 2018

Helix, Nectar and RDS partnership

Aligning research infrastructure Hellix is partnering with Nectar and RDS to deliver transformation in the research sector. Read more

The project

Hellix-Nectar-RDS News is a great source of news, events and data jobs - sent direct to your inbox every fortnight. Sign up now to get the next edition.

Don't miss out!

Deliverables

Hellix-Nectar-RDS News is a great source of news, events and data jobs - sent direct to your inbox every fortnight. Sign up now to get the next edition.

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Έρευνα



Συνεργάτες











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Maria Ikonomou Senior Data Biologist @UATHENS PhD in Ecology. Knowledge seeker and food lover.	+ Mountain herbs
 @AthenaRC Athens, Greece maria@uathena.com http://mariaathena.com Joined on Jan 6, 2019 	and traditional medicine 2 repositories 2 PUBS 5 DATA
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About

Αρχική

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Data
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Συνεργάτες











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Discover and share open scientific publications

HELIX Architecture **Publications**

Search for Publications

- Harvested from EU-wide institutional, thematic, or ad-hoc repositories
- Provide publications published from Greek S&T organizations
- OAI-PMH v2.0, OAI-DC
- Value added services (under development/testing)
 - Infer data from publications (link data with pubs)
 - Analytics & KPIs
- OA Training & support





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Hellenic Data Service

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Featured Publications		
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Institutution

Amann, M. (2017) Cited By 2081

PUBLICATION: LAST REVISION: LANGUAGE: TYPE:

PUBLICATION

Το σύνολο γεωχωρικών δεδομένων απεικονίζει μικρούς νησιωτικούς υγροτόπους έκτασης μικρότερης των 80 στρεμμάτων, όπως εγκρίθηκαν από τον Υπουργό Περιβάλλοντος Ενέργειας και Κλιματικής Αλλαγής, με την εξουσιοδότηση του νόμου για τη Διατήρηση της Βιοποικιλότητας (ν.3937/2011) και μετά από επαρκή επιστημονική τεκμηρίωση. Το σχέδιο Προεδρικού Διατάγματος για την προστασία των μικρών νησιωτικών υγροτόπων με τον τίτλο: «Έγκριση καταλόγου μικρών νησιωτικών υγροτόπων και καθορισμός όρων και περιορισμών για την προστασία και ανάδειξη των μικρών παράκτιων υγροτόπων που περιλαμβάνονται σε αυτόν» υπεγράφη από τον Υπουργό την 1η Φεβρουαρίου και προωθηθηκε στο Συμβούλιο Επικρατείας για επεξεργασία.

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Applied System Analysis



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Find, view, and use open scientific data

HELIX Architecture Data

CKAN-based Data Catalogue & Repository extended via multiple plugins

- Core CKAN v2.8 (started from v2.2, soon will port to v3.0)
- Plugins: CKAN + PublicaMundi (metadata, geo) + HELIX (metadata/workflow)
- Custom roles/profiles/organization structure

Core CKAN services & HELIX-specific services

• Search, view, visualize, download

• Data management

- Dataset upload (files) open to all publishers (size-limited, admin QA & sanitization)
- Multiple **replication** policies for harvested datasets
- Automated independent and asynchronous data ingestion policies (files to data)





HELIX)AIA

HELIX Architecture /// Data **Data Services**

- Core Metadata and Standard Schemas
 - **DataCite-based** schema (default, common, simple)
 - Support for domain-specific metadata schemas (e.g., ISO 19131)
 - Upload/harvest (e.g., INSPIRE or Public Data catalogues)
 - Extensible programmatic homogenization/mapping (to Core), Ul generation (editor) and on-the-fly transformations (all metadata files available)
- Personal data collections (check later, send to others, use in Lab)
- Datasets linked with Data Services (how/where to use) & Pubs (manual & automated via OpenAIRE)
- User hierarchies/rights (organization, curators, authors)
- Flexible **DMP** support (confidential, embargo)



HELIX Architecture /// Data **Data as a Service**

- **Data catalogue & repository** provided as **a** Service to Research Organizations, Scientific Infrastructures, Domain-specific communities, Government/NGOs
 - Follow the data and the users (high-value data, large user groups) and bring the services closer to their **needs** (e.g., domain-specific schemas and services)
 - Low-cost, low-effort, inclusive institutional data catalogues/repos with integrated OA support & DMP facilities
- Sub-domain in HELIX (group)
- White-labelling



Data as a Service

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Featured Data sets

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SHAPEFILE WMS WFS

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Latest Data Sets

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 ΕΠΙΧΕΙΡΗΣΕΙΣ

 A Brief History Of Time

 ΓΕΝΙΚΗ ΓΡΑΜΜΑΤΕΙΑ

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Πριν 8 ημέρες @ ΕΠΙΧΕΙΡΗΣΕΙΣ Θέσεις Αρχαίων Μνημείων Γενική Γραμματεία Shapefile WMS

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SHAPEFILE WMS WFS



About

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Data Publica Lab

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Institution

Ministry of Environment, Energy and Climate Change

Topics

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PUBLICATION: LAST REVISION:

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DATA AND RESOURCES



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EMAIL	info@geodata.gov.gr					
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helix

Learn, experiment, and build with data

HELIX Architecture Lab

Open-ended collection of independent services and applications for experimenting and using data

- No interdependencies or single point of failure
- Fast and simple to replace/extend services in operation
- Service portfolio constantly expanding, with varying TRL/access levels
- Replicate/expand the industry emerging paradigms (e.g., Azure, Google)
- All have automated & configurable access to the repository's data
 - Data available as files or databases/data processing frameworks
 - Flexible data availability policies per type/data set (e.g., depending on size, popularity, importance, domain, resource-utilization)





HELIX

HELIX Architecture /// Lab **Data Science Notebooks**

- Jupyter Lab/Hub (open beta)
 - **Tiered** kernel/resource per user (from R, to HPC)
 - **Repository** data available in user's notebooks (my data collection; minimize time/effort to discover & use data)
 - Support for under/post-grad courses (share data/exercises) and **research teams** (collaborative editing)
 - Constantly expanded with additional facilities & services to support **Data Science** and targeted domain needs
- **Apache Zeppelin** (invitational beta)
 - Notebook-like facility for **Apache Spark** (Java/Scala)
 - Dedicated clusters for **Big Data** experimentation & benchmarking
- **Reproducible & Operational research** (TBA)
 - Containerized algorithms & data
 - Deployment-ready



Interactive computing

Scalable Computing







Βρείτε δεδομένα συμβατά με Jupyter

Πηγές από 100+ κυβερνητικές υπηρεσίες και ερευνητικά ιδρύματα

Interactive coding in your browser

Free, in the cloud, powered by Jupyter



Introduction to Python

Learn the basics of Python 3 in Azure Notebooks. Learn Python syntax, standard data types, as well as how to write a simple program.



Introduction to R

Get a brief introduction to charting and graphing capabilities of R in the Jupyter Notebook. You will learn how to make line charts, pie charts and scatter plots.

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About

Αρχική

Το έργο

Επικοινωνία

Όροι χρήσης

Έρευνα

Data
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Θεματικέ
Οργανισμ

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Powerful Languages

Azure Notebooks provides execution environments for Python 2, Python 3, F#, and R.

Use the languages of Data Science

Numerous Charting Libraries

0

Quickly visualize your data and results using plotting libraries such as ggplot, matplotlib, bokeh, and seaborn.

Inline Graphs



Introduction to F#

Get a brief introduction to using F# in the Jupyter Notebook.

Introduction to Python 3

Learn the basics of Python 3 in Azure Notebooks. Learn Python syntax, standard data types, as well as how to write a simple program.

Συνεργάτες













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About

Αρχική	
Το έργο	
Επικοινωνία	
Όροι χρήσης	

Data

Lab

Publications

Θεματικές

Οργανισμοί

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Συνεργάτες Έρευνα











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HELIX Architecture /// Lab **Other Services**

Interactive Data Services/widgets (evaluate & use)

- Presentational (tables, charts, maps) for tabular data •
- File transformations (schemas/formats, CRS)
- **End-points & APIs** (for third system/apps)
 - OGC Services for geospatial (Catalogue, WMS, WFS, WPS-experimental)
 - Linked Open Data (SPARQL, GeoSPARQL end-points)
 - JavaScript Data API (simple filter/SQL-type queries over tabular data)
 - JavaScript Mapping API (custom standalone/embeddable maps)



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Publish Data Data stories Share code Data curricula Partners

Universities Research Centers Scientific Infrastructures Digital Libraries Communities

Data catalogue & repository Harvest **Data Science** OA & DMP training

We can provide you with



Use Data Publish Data Data stories Build Spread the word



Scientist Researcher Innovator **Citizen Scientist** Student

We can provide you with

High-value data Interactive Computing Scalable Computing



Publish Data Promote OA Align policies Support expansion



Public Sector Grant manager NGO

Open Data Citizen engagement **Data Services OA Policy**

We can provide you with



Discover & Harness Data **Build data**driven services **Contribute data** & stories



Industry Innovators Startups

We can provide you with

High-value data Computing Infrastructures **Data Science** Training



















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Learn, experiment, and build with data