

Workflow Based Tools for Integrated Spatiotemporal Research

Wendy Guan, Lingbo Liu & Xiaokang Fu Center for Geographic Analysis, Harvard University Shuming Bao Future Data Lab

The Era of the 4th Industrial Revolution

Artificial Intelligence

Bioinformatics

Digital Twins

Cloud Computing

Internet of Things

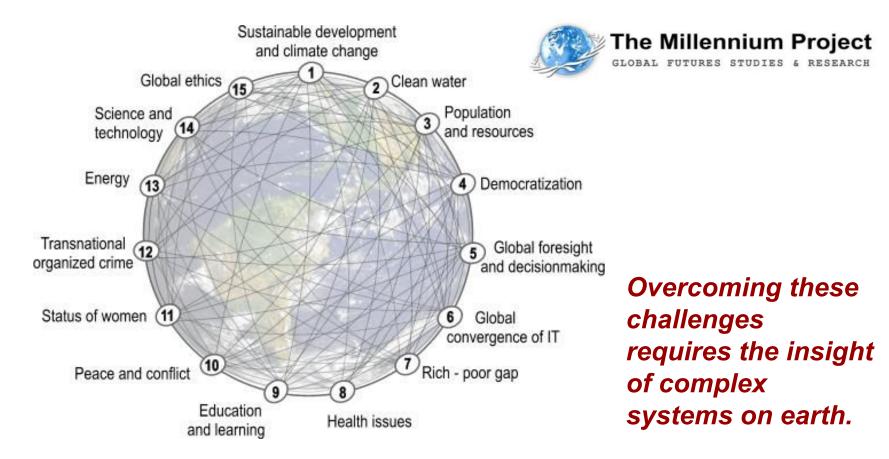
Robotics

Block Chain

Big Data

Geolocation becomes ubiquitous, place and time are embedded

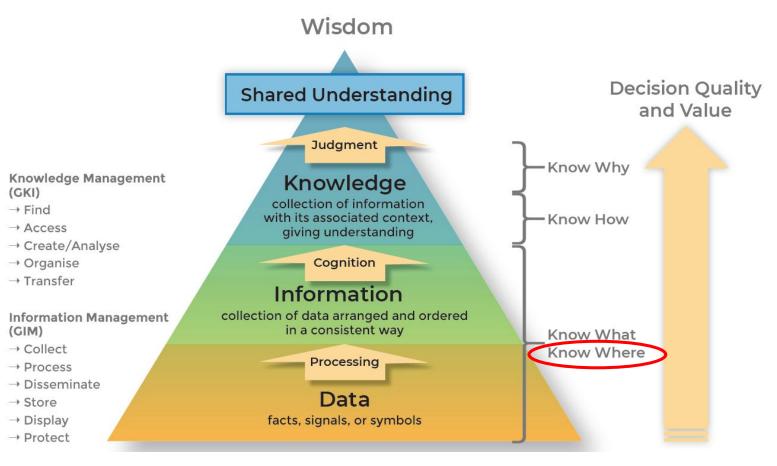
The 15 Global Challenges



Geospatial & Temporal Data Are Everywhere

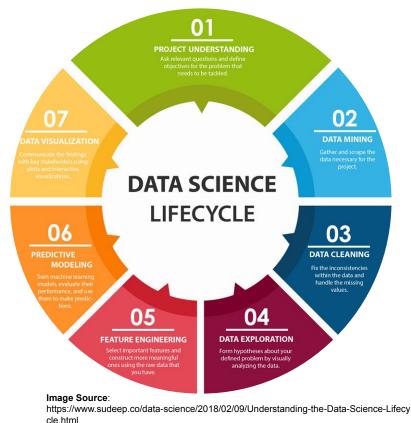


How to Solve Problems with It?



Spatial Data Science Research Needs

- Easy data access, especially for frequently updated data
- Heterogeneous data integration
- Efficient data analysis and visualization
- Access to high performance computing, especially for spatiotemporal big data
- Reproducible and replicable results
- Collaboration among researchers from different fields with different levels of skills



Spatial Data Services

gectdata.gov

- Data & metadata download
- Query and pre-view
- Shared data editing
- **Online Analysis**
- Publishing
- Preservation



your one stop for federal,

state & local geographic data

Home About Us Help

Oceans

0 0

Coasts

Challenges in Spatial Data Services

- High development cost and slow implementation
- High maintenance requiring professional skills
- Lack of flexibility for customizable research inquiries and changing demands
- Barriers when sharing with researchers from different fields at different knowledge and skill levels



https://aub.edu.lb.libguides.com/data_services/home

Challenges in Spatiotemporal Research

• Data Barriers

- Discovery (sources unknown, scattered)
- Accessibility (restricted, sensitive, licensed)
- Big data (volume, variety, velocity, veracity)

Tool Barriers

- License
- Operating environment
- Computing power
- Maintenance (security, version updates)

Implementation Barriers

- Methodology (reproducible, replicable, generalizable)
- Technology (collaborators' varying backgrounds & skills)
- Applications (efficiency, effectiveness, scalability)



About 74% re-executions failed

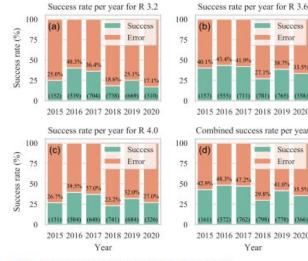


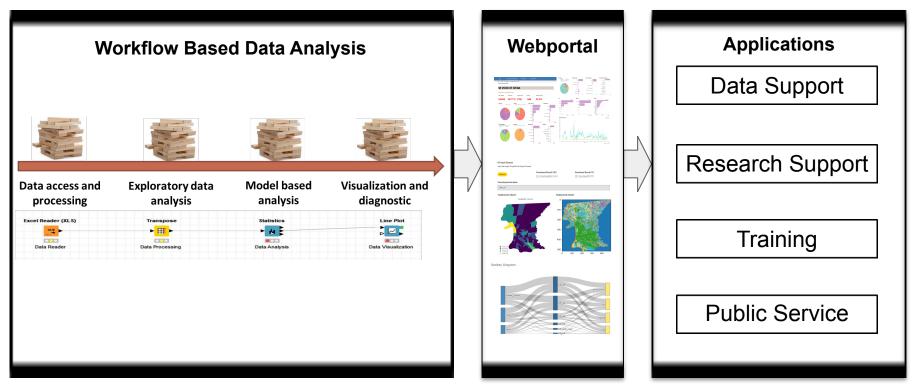


Image source:

https://www.nature.com/articles/s41597-022-01143-6.pdf?origin=ppub

Our Approach - the Spatial Data Lab (SDL)

An Integrated Solution for Spatiotemporal Research, Applications, Training, and Services



About the Spatial Data Lab Project



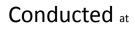


NSF IUCRC Spatiotemporal Innovation Center(STC)











Geographic Analysis

Harvard University

Hosted at





Build a platform for effective and collaborative spatiotemporal research



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OBJECTIVES

- Cloud-based spatial data integration
- 00 Executable quantitative analysis tools

GOAL

- Workflow-based case studies
- Education and training

MAJOR PRODUCTS



Spatial data services

Spatial data analysis platform and tools

Workflow based case studies

Training programs & research papers

Current and Previous Work

- Data collection and sharing: COVID-19 Data, Mobility Data, available via Dataverse API
- Research applications and support: Workflow-based Case studies (60+ case studies based on peer-reviewed publications; team members published 30+ papers)
- Workbooks for textbook and teaching lab: 30+ case studies (theory, models, and tools)
- **Training workshops:** 30+ training workshops on research data collections, COVID-19 studies, and spatiotemporal studies on various topics



KNIME as the Foundation for the SDL

- Open source (http://knime.com)
- End to End Data Science, enabling automation from data input, manipulation, visualization, mining, reporting and web-based services
- Codeless, Visual and workflow-based Programming with 2000⁺ built-in nodes
- Extension for coding snippet of Python/ R /Java
- Integrated environment for desktop version and server version
- Workflow-driven Web Portal service for easy-to-use, interoperable, and executable data modelling and reporting



KNIME Analytics Platform

KNIME Analytics Platform is the free, open-source software for creating data science.



KNIME Server

KNIME Server is the commercial solution for productionizing data science.

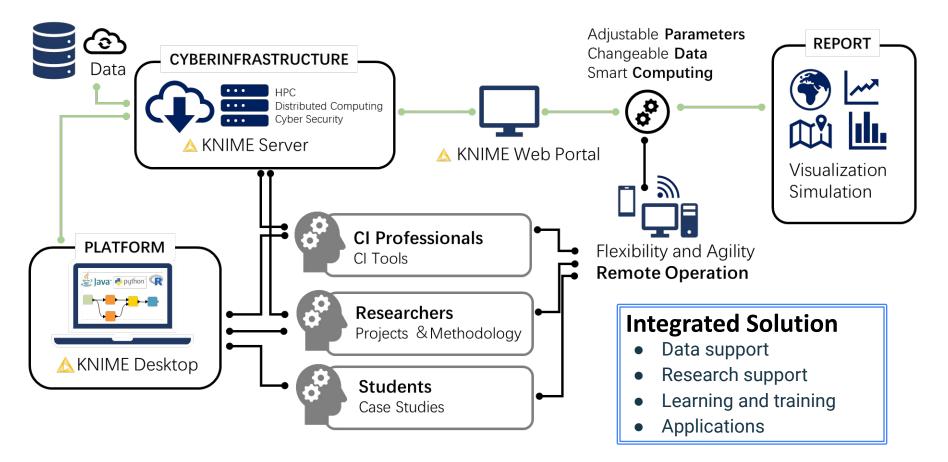
End to End Data Science

At KNIME, we build software to create and productionize data science using one easy and intuitive environment, enabling every stakeholder in the data science process to focus on what they do best.



Image Source: https://www.knime.com/

The KNIME-based SDL Architecture Overview



The Integrated Solution Supporting Spatiotemporal Research

Data Support

- Out-of-Box Application without Coding
- Adjustability on Key Parameters in Modelling
- Reproducible with New Data

Research Support

- Integration of Local and Server for Sharing
- Data Privacy Protection
- Expandable Workflows

Learning & Training Support

- Ubiquitous Access
- Flexible Skill Requirements
- Tailorable Modules

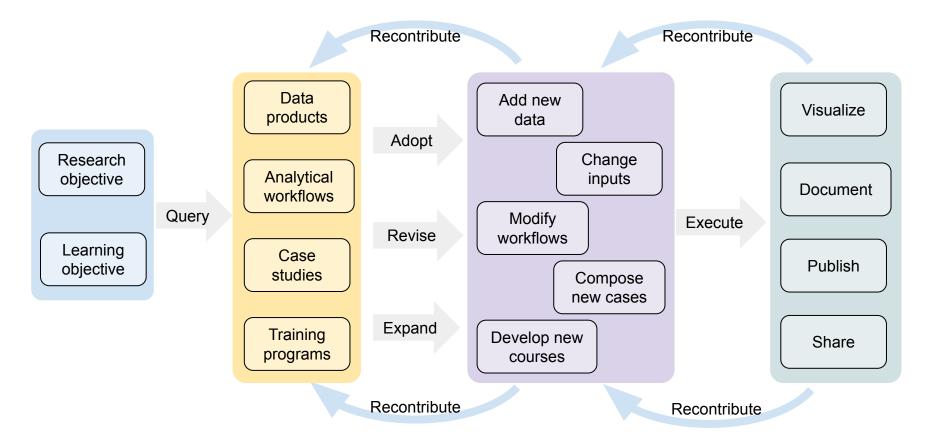
Application Support

- Data and Report Automation
- Easy Maintenance
- Interactive Usage on Web Page

Advantages in Spatial Data Service

- Efficient development and implementation of customized research services
- Easy maintenance of the system without special skills
- Great flexibility on expansion for changing demands
- Low entry cost for learning and applications
- Promoting reproduction and replication of quantitative research
- Facilitating multidisciplinary and cross-organizational collaboration

The Re-engineered Research Lifecycle in SDL



Geospatial Analytics Extension for KNIME

- Release in December 2022
- Easy installation via drag and drop from KNIME Community Hub
 - Batteries included no additional setup and installation steps required
- Support for most common vector data, e.g. points, lines, polygons, collections
- Joined development by the Center for Geographic Analysis at Harvard and KNIME
- We will continue to work together to add more functionality and improve the usability
- Feedback is always welcome via KNIME Forum or Github*

*<u>https://github.com/spatial-data-lab/knime-geospatial-extension</u>

End to End Geospatial Analytics



Spatial IO



Spatial Calculation



Spatial Manipulation



Spatial Transformation



Spatial Conversion



Spatial Visualization



Exploratory Spatial Data Analysis



Spatial Modelling



Location Analysis



Open Datasets

The Application Cases of Geospatial Analytics



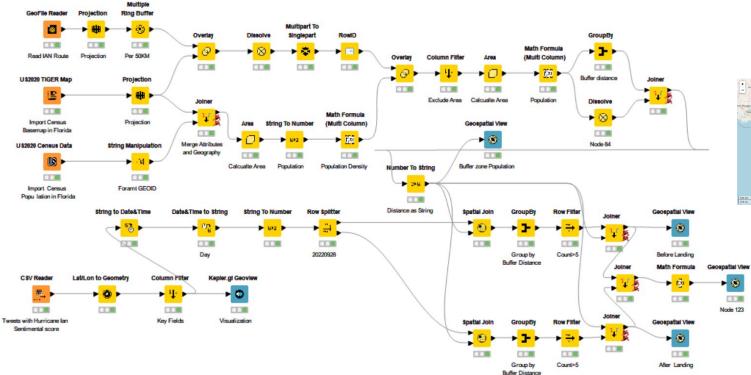
https://www.youtube.com/playlist?list=PLnFUy1r9kH-20dWQGVKKiUAOlbPGxyBUv

C KNIME GEOSPATIAL ANALYTICS

Impact of Hurricane Ian

Hurricane IAN- Impact Analysis

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Sponsors and Contributing Open Source Projects



NSF IUCRC Spatiotemporal Innovation Center(STC)



National Science Foundation



Lea

Future Data Lab











The Python Spatial Analysis Library for open source, cross platform Geospatial Data Science



Charting the Future of the SDL

Sustainable Development:

- Augment the limited team members housed at Harvard
- Search for mechanisms for continued expansion
- Cultivate the communities of development and applications
- Adapt to and evolve with new technology and methodology

Global collaboration:

- Data collection and service
- Tool development
- Workflow-based case studies development
- Training on spatial data analysis
- Research applications

The Summer Workshop on Spatiotemporal Innovation

Date: Mon - Fri, Jul 10 to Jul 14, 9:00am - 5:00pm

Location: 1730 Cambridge Street, Cambridge MA 02138, Cambridge, MA

Topics (5 days) :

- 1. An Introduction to Workflow-Based Tools: KNIME Analytics Platform, KNIME Server and KNIME WebPortal
- 2. Open Data Access, Integration, Online Data Sharing and Visualization
- 3. Spatiotemporal Analysis with Geospatial Analytics for KNIME
- 4. Artificial Intelligence Analysis for Spatiotemporal Data
- 5. Workflow-based Remote Sensing Analysis
- 6. Business Intelligence Analysis with Spatiotemporal Data
- 7. Spatiotemporal Analysis with Social Media Data
- 8. Spatiotemporal Analysis with Healthcare Data
- 9. Introduction to Affiliate Labs on Spatiotemporal Innovation
- 10. Discussion on Future Collaborations









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Acknowledgement: This work is partially funded by NSF #1841403