

Informing Climate Change and Sustainable Development Policies with Integrated Data

BILBAO. SPAIN 10-14 JUNE 2024 #UNBigData2024

Measuring Resilience: Big Data Approaches to Climate Change Risk Assessment ARIES application to measure physical and transition risks

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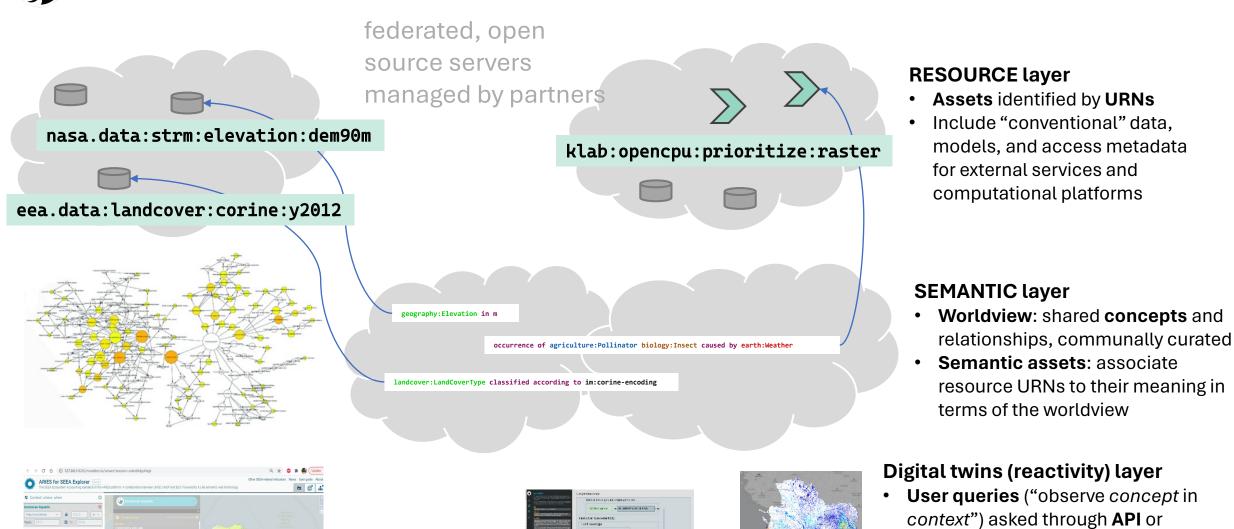




ARIES: Artificial Intelligence for Environment & Sustainability

#1	It is a modelling technology , rather than a collection of models or specific program/application,
#2	It is an AI modeller , based on machine reasoning , a less known branch of AI;
#3	It defines a variety of data, models and the relationships between them using consistent and uniform terms . This allows different data and models to be used together, depending on which data and models are "most appropriate" for the context set by the user;
#4	It uses AI to determine the "most appropriate" data and models for users' requests.
Reaso algori	

A semantics-driven, Al-assisted model and data federation



im:Net monetary value of ecology:Pollination

- applications All assembles the best-case
- AI assembles the best-case algorithm to produce reactive **observations**
- **Behaviors** can be specified and triggered



Potential use cases for ARIES: Measure physical and transition risks





Spatial economic valuation of ecosystem services Conservation planning



Spatial policy planning (derived from EO data) Forecasting changes in ecosystem service provisioning

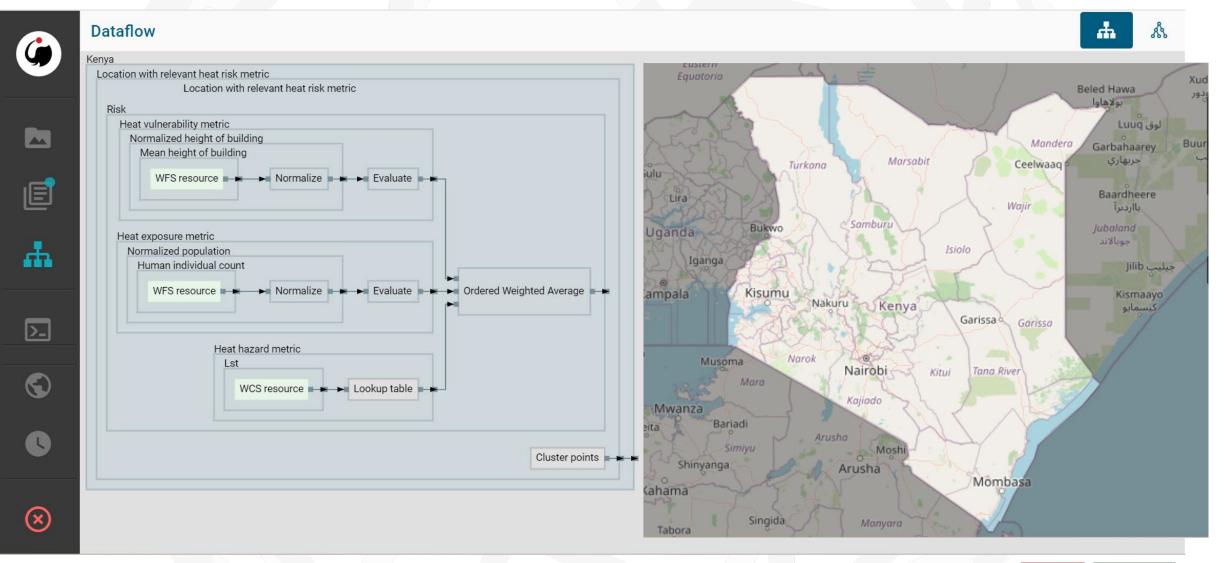


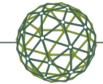
Natural capital accounting



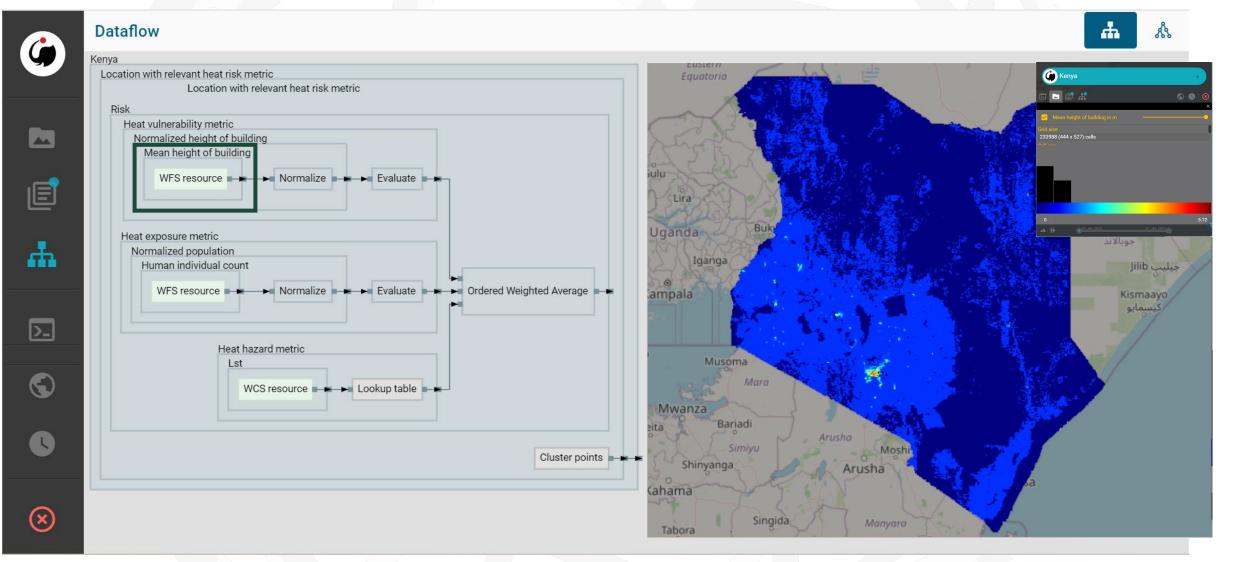
Measure physical and transition risks

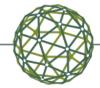
Kenyan context



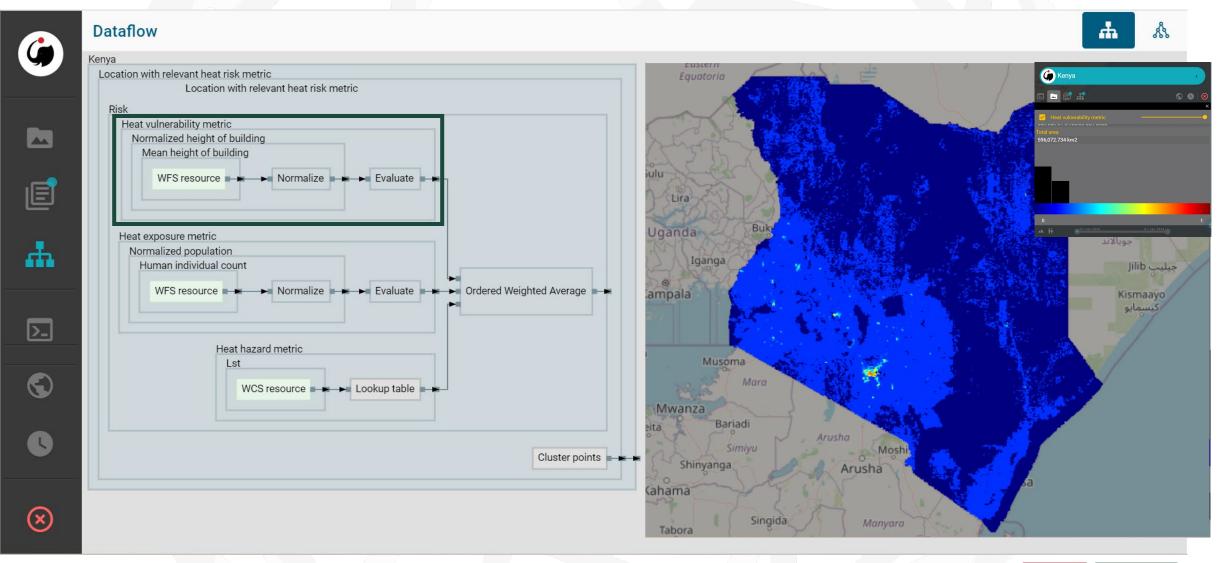


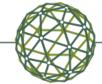
Mean height of building



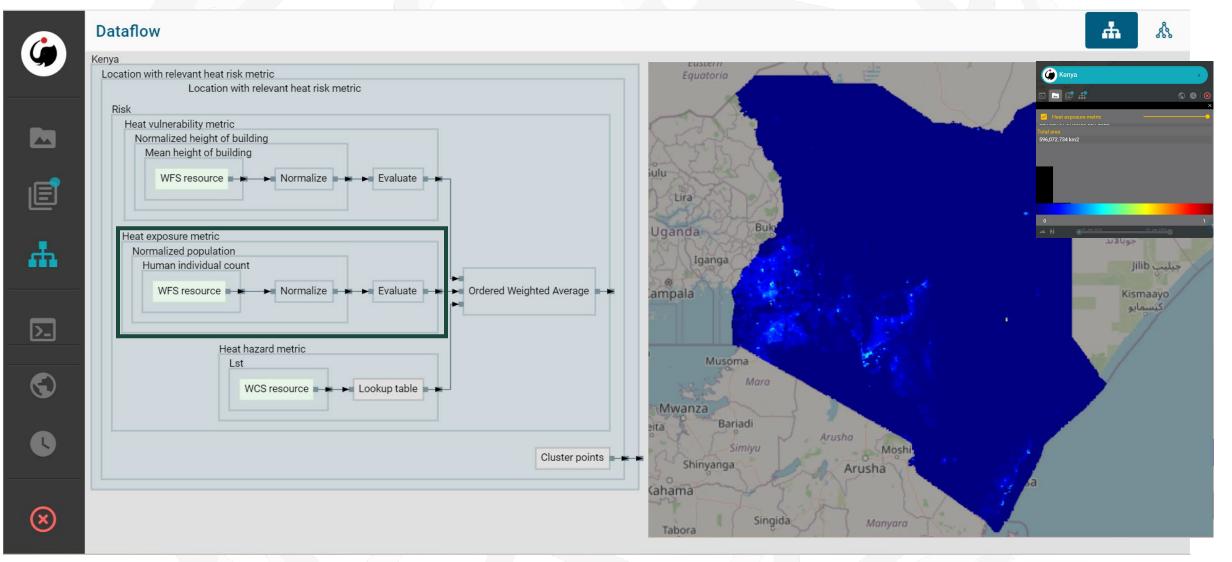


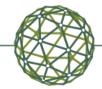
Heat vulnerability metric



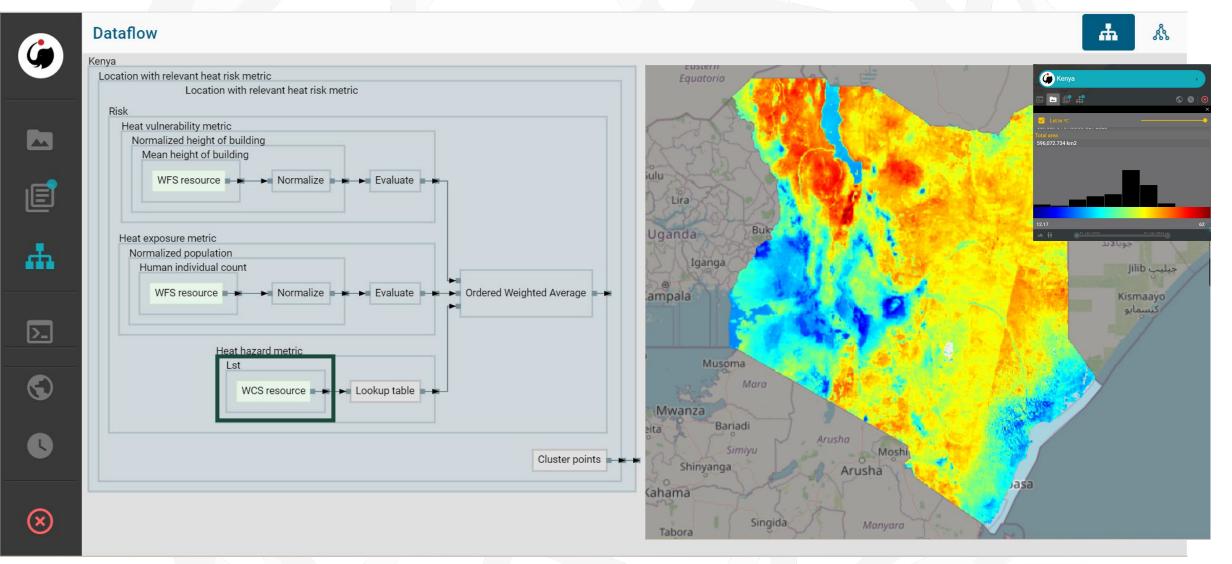


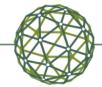
Heat exposure metric



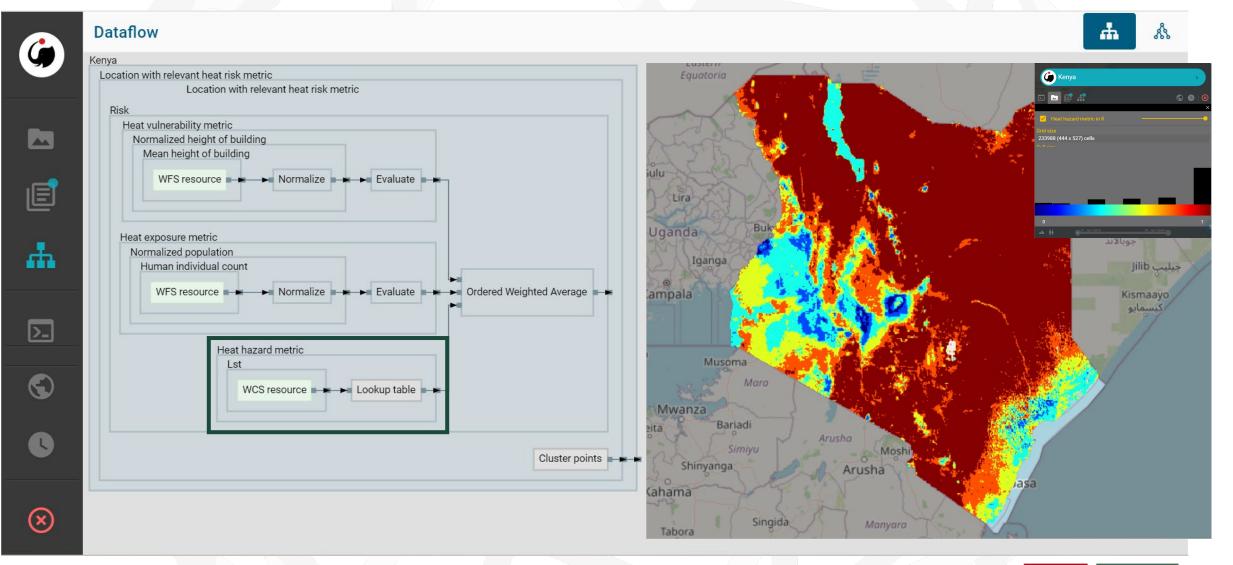


Land surface temperature



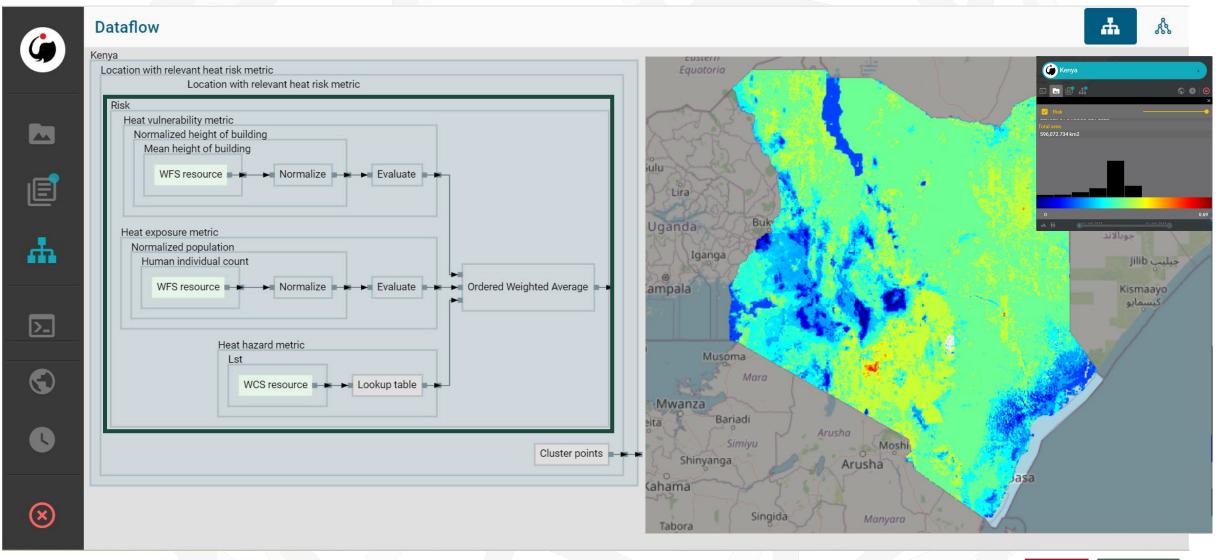


Land surface temperature



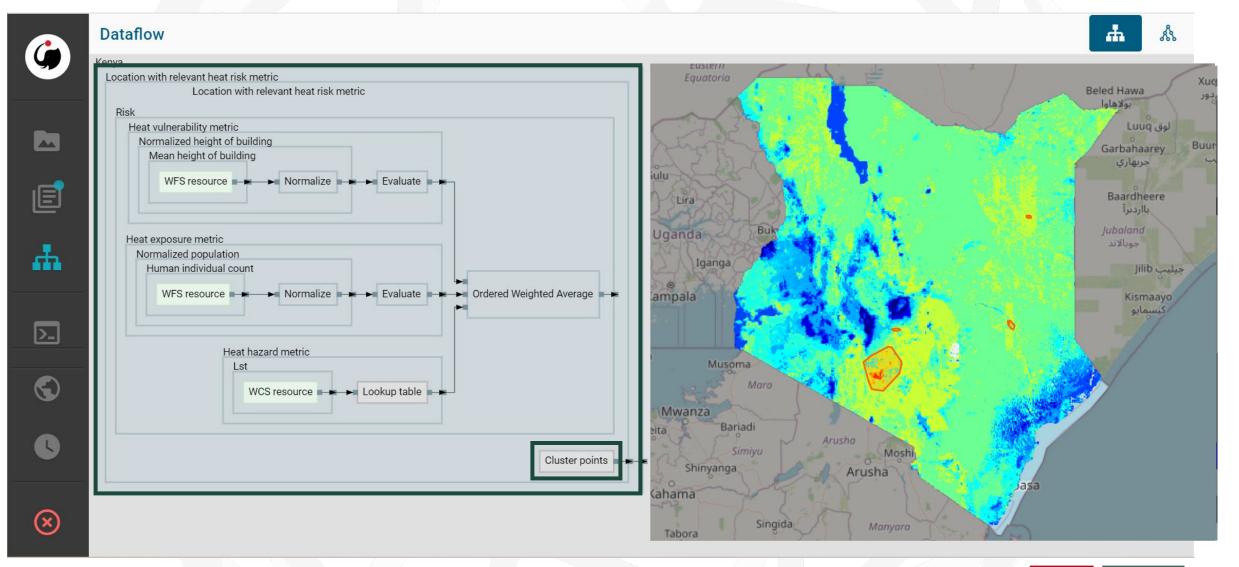


Risk



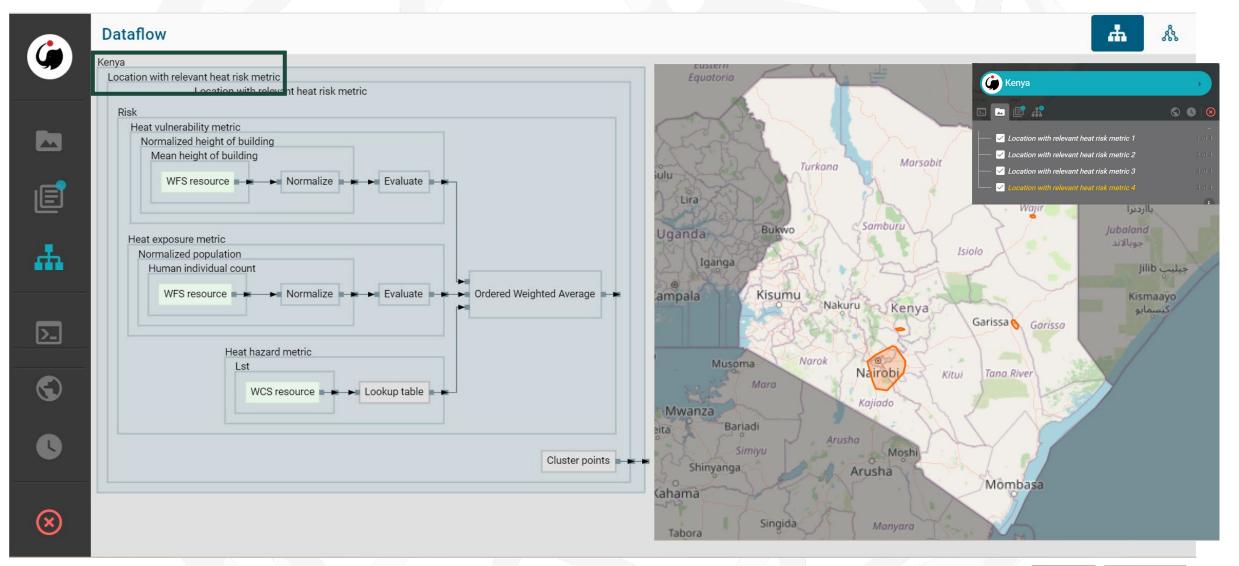


Locations with relevant heat risk metric (>3/5 of max risk)





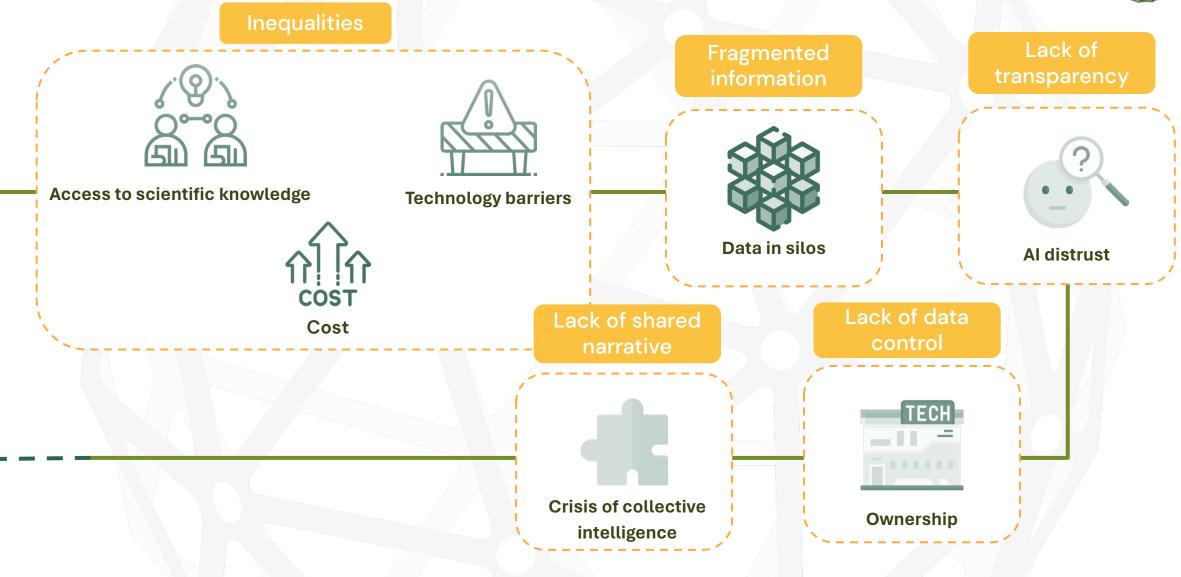
Locations with relevant heat risk metric (>3/5 of max risk)







Challenges of science-driven decision making > Complex systems, complex issues, fragmented knowledge



Integrated Modeling



is a practice meant to maximize the value of scientific information by ensuring its modularity, reusability, interoperability and traceability throughout the scientific process.



4. Adaptive contextualization

Adaptively incorporate the best-available knowledge, from curated global public datasets to "big data" to user-provided data.



5. Delivery

Adopt shared, non-ambiguous semantics in the implementation, documentation and dissemination of products.



6. Tracking Track quality, uncertainty and provenance throughout modelling workflows.

Balbi et al. The global environmental agenda urgently needs a semantic web of knowledge. Environmental Evidence 11, 5 (2022). doi:10.1186/s13750-022-00258

1. Linkage Combine independently produced scientific products into workflows that would be too complex for individual humans to conceive, validate and navigate.



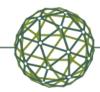
2. Integration

Integrate different modelling paradigms from simple (e.g., deterministic and probabilistic models) to complex approaches (e.g., agent-based and networks) depending on context and scale.



3. Rescaling

Rescale smartly across scales, from local to global, promoting adaptive solutions that are automatically customized to the scale of observation.



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Relevance of semantics and ontologies for ARIES

 Unambiguous identification of concepts, data, models and results. **Description** of real environmental scenarios.

• Interoperability of data and models.

 Transparency with users in the methodologies applied and results obtained.

Transformation of Knowledge to Knowledge – an insight into modelling in

