#### Monitoring Bio- Geodiversity and Ecosystem health by Traits, Remote Sensing (RS) and Data Science approaches



Spaceborne



Airborne





UAV - Drone



Camera trap



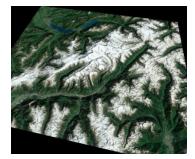
Wireless-Sensor-Network (WSN)

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#### PD Dr. Angela Lausch

Helmholtz Centre for Environmental Research – UFZ, Germany Angela.Lausch@ufz.de

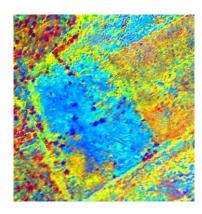
### How can RS measure, assess Bio-and Geodiversity & and ESS?

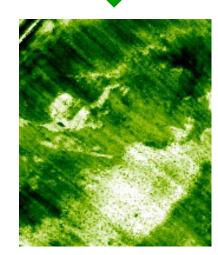


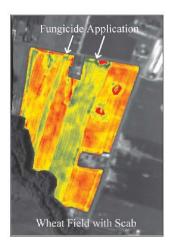












Spaceborne



Airborne





UAV - Drone



Camera trap



Wireless-Sensor-Network (WSN)



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#### **Trait concept of species – Indicators / Filters of stress**

"Ecologists are increasingly looking at traits - rather than species - to measure the health of ecosystems"

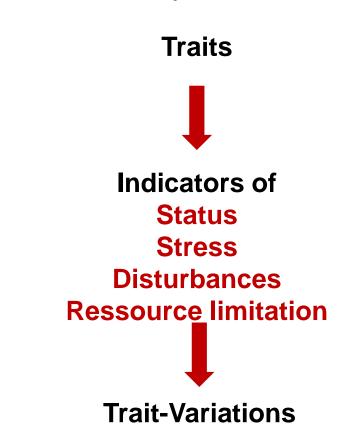
# The biodiversity revolution

Ecologists are increasingly looking at truits - rather than species - to measure the health of ecosystems.

#### BY BACKEL CERMINET

monett Dufy was about 5 metres under only on the number of species





Cernansky, R. Biodiversity moves beyond counting species. *Nature* 2017, 546, 22–24

**Approach: Trait concept of species** 

# Species traits allowed us to go a "complete new way in understanding of fundamental questions of biodiversity"

(Green et al., 2008)

Traits help us to understand:

"Why organisms live where they do and how they will respond to environmental change"

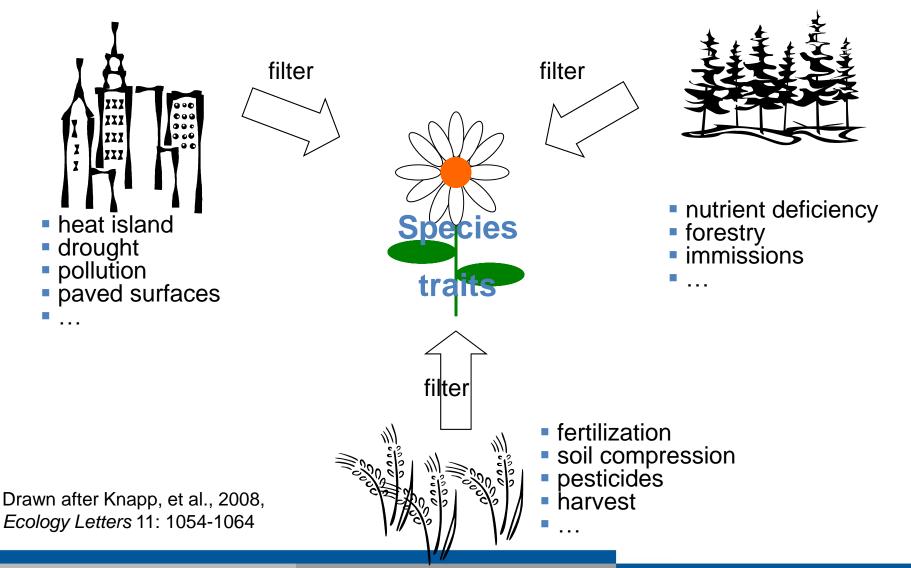
(Green et al., 2008)

And how they interact to different stressors, disturbances, resource limitations and drivers

Green, J.L.J.L., et al., 2008. Microbial biogeography: from taxonomy to traits. Science, 320, 1039–1043. doi:10.1126/science.1153475

#### **Approach:** "In-situ-species traits" - concept

# **Traits** = Filters of status, stress, processes, disturbances and resource limitations



### **Approach: Remote Sensing – Trait concept**

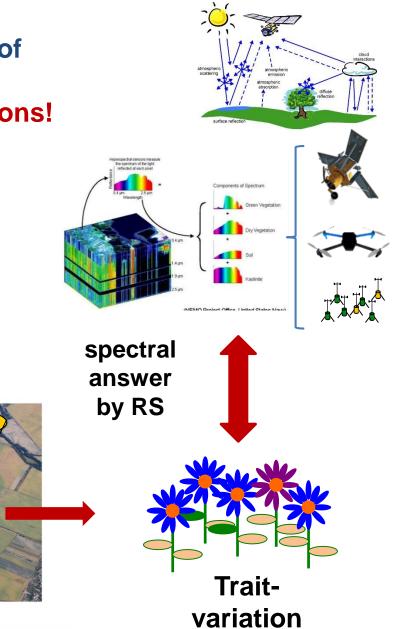
- RS record "Traits and Trait variations" of surface, vegetation, soil, water ...
- > Bio-and Geodiversity and their interactions!

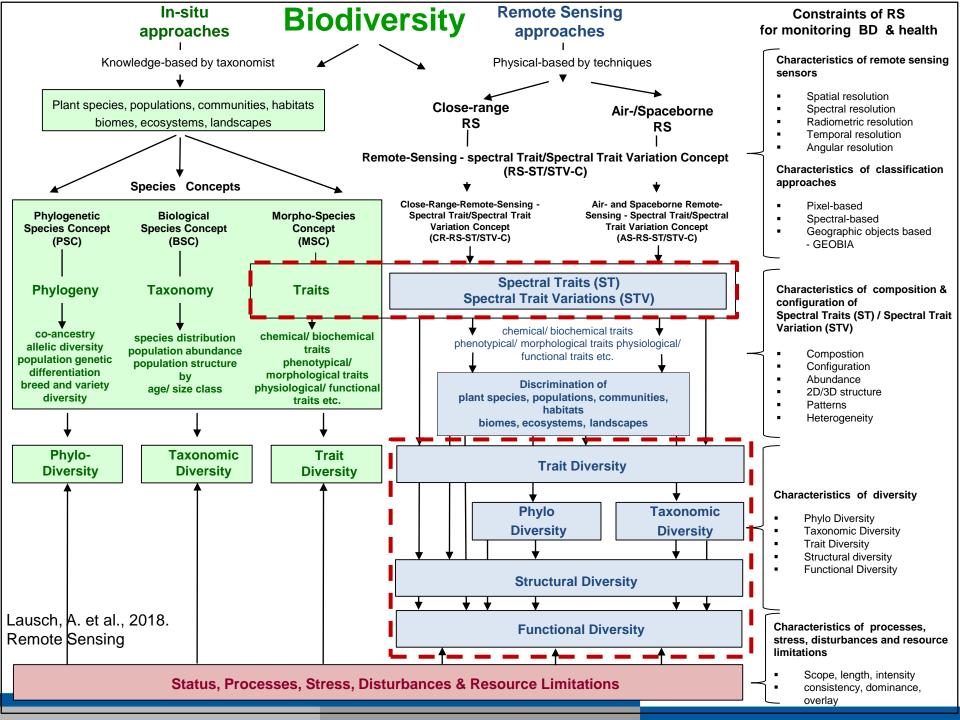
**Processes** 

- Spectral answer, is a reaction on
  - status, process,
  - disturbances,

**Traits** 

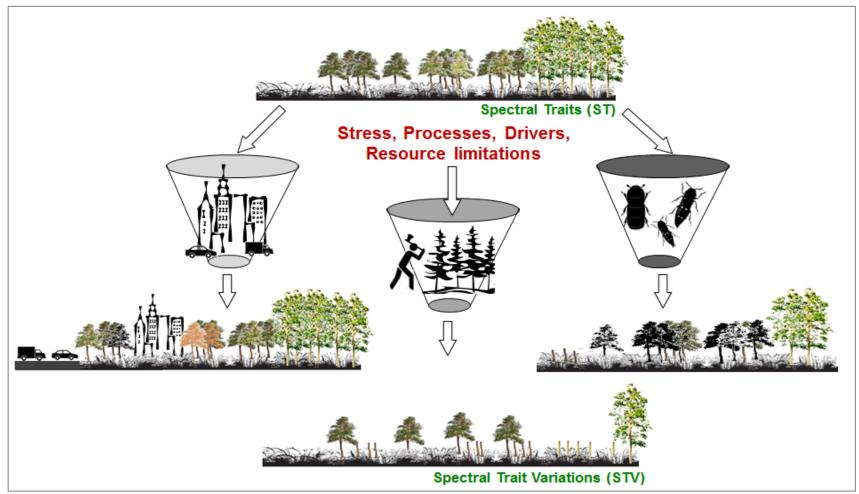
- ressource limitations
- pattern process interaction



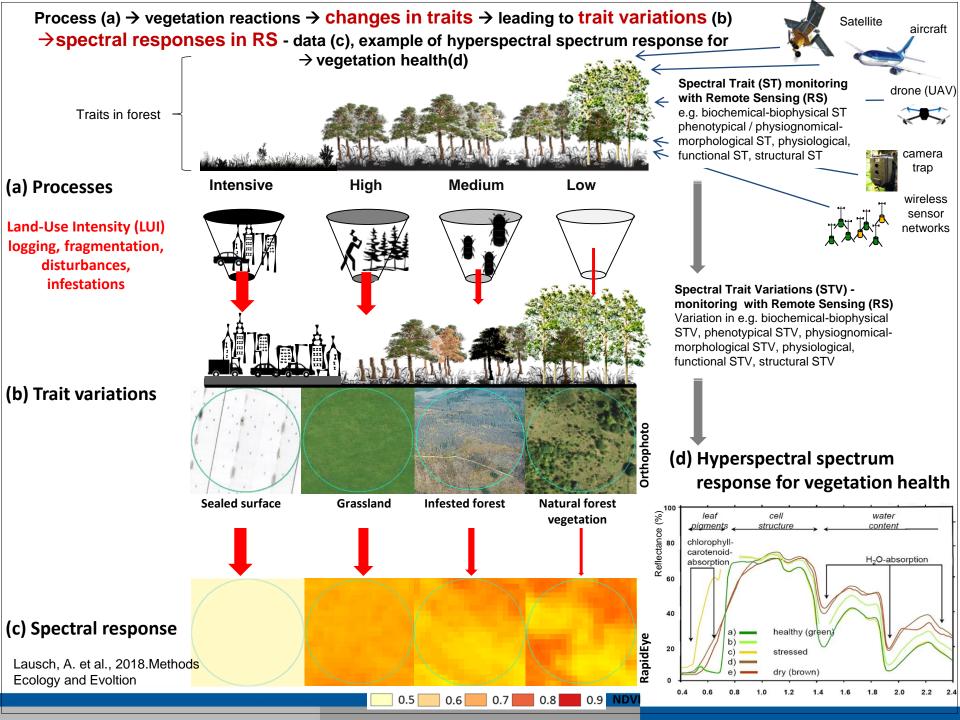


### **Trait concept – Indicators / Filters of stress**

# **Traits** = Filters for stress, processes, disturbances and resource limitations



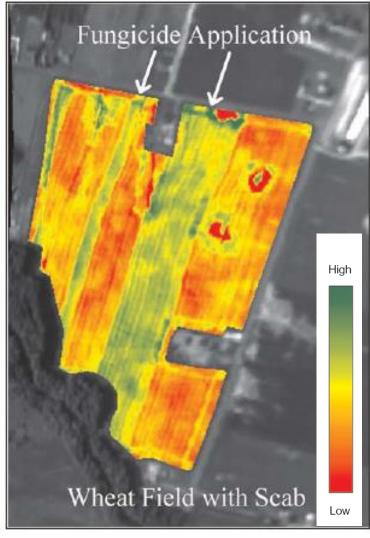
Lausch, A., et. al., 2016. Understanding Forest Health with Remote Sensing -Part I-A Review of Spectral Traits, Processes and Remote-Sensing Characteristics. Remote Sens. 2016, Vol. 8, Page 1029 8, 1029. doi:10.3390/RS8121029

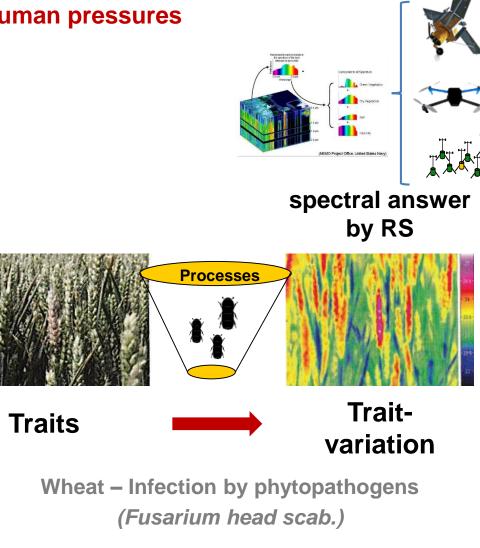


#### **Trait concept - Understanding - Processes – Pattern – Interaction**

Monitoring and quantification of fungicid, pesticide applications

- Healthy status of vegetation
- Changes of plant traits based on human pressures





### **Trait concept of species -Traits - Animals**

#### Opinion

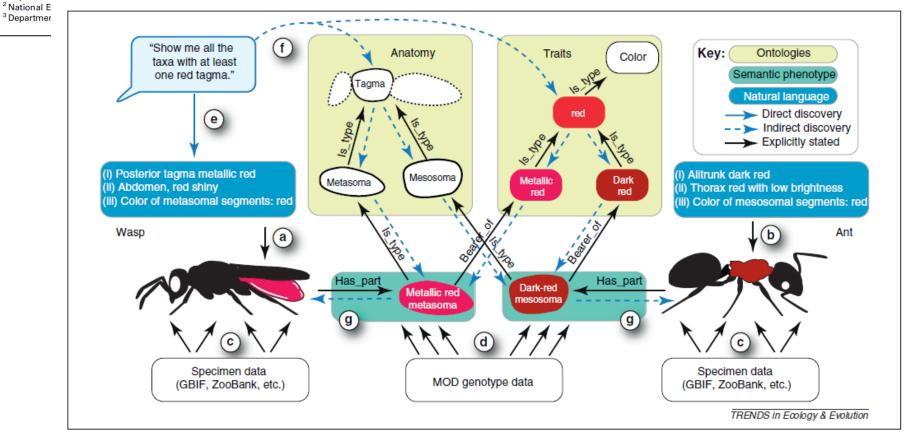
Special Issue: Ecological and evolutionary informatics

#### Time to change how we describe biodiversity

Andrew R. Deans<sup>1</sup>, Matthew J. Yoder<sup>1</sup> and James P. Balhoff<sup>2,3</sup>

<sup>1</sup>Department of Entomology North Carolina State University Raleigh NC 27695 USA

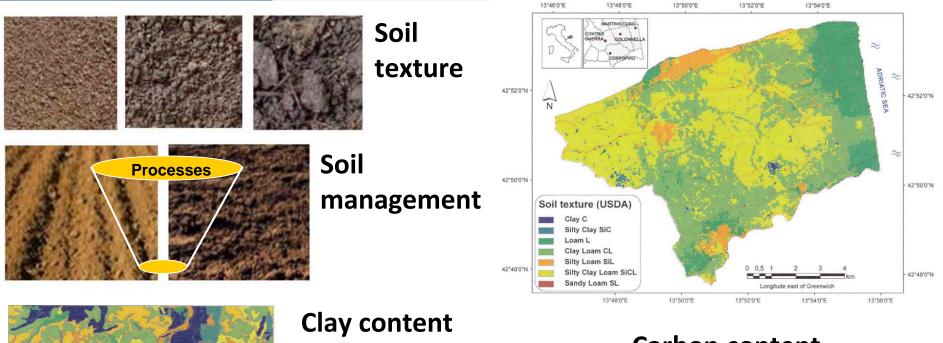


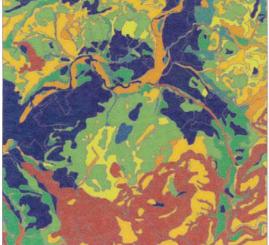


Deans, A.R., Yoder, M.J., Balhoff, J.P., 2012. Time to change how we describe biodiversity. Trends Ecol. Evol. 27, 78-84.

### **Trait concept – Traits of Geodiversity**

#### Soil texture

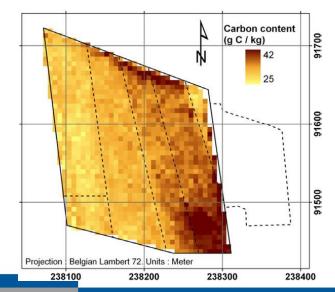




Clay	content %
22	5 - 10
5-5	10 - 15
Sec.	15 - 20
100	20 - 25
100	25 - 30
	30 - 35
	35 - 40
	40 - 45
	45 - 50
	50 - 55
	55 - 60

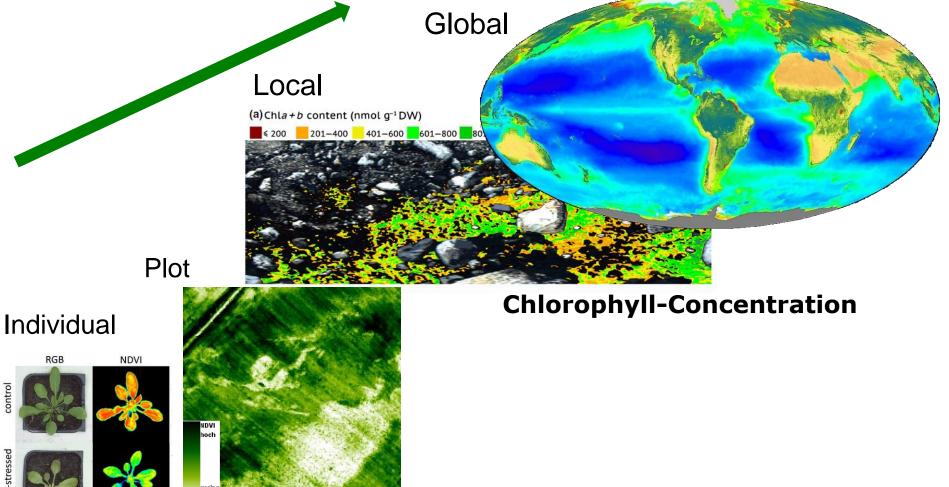
Wulf, H., Mulder, T., Schaepman, M. E., Keller, A., & Jörg, P. (2014). Remote Sensing of Soils. *Zurich, Switzerland*, 71.

#### **Carbon content**



### **Trait concept - Understanding - Processes – Pattern – Interaction**

- Traits  $\rightarrow$  Exist on all spatial and temporal scales
- Imortant: Linking of traits between scales

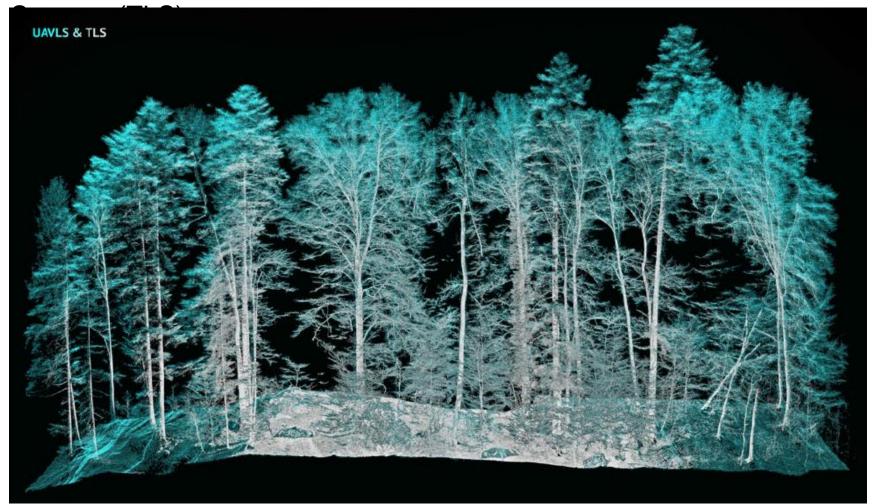


Malenovsky et al., 2015; https://upload.wikimedia.org

0.49 0.62 0.76 0.9

### **Ecosystem health by RS 2-4D Structural Diversity**

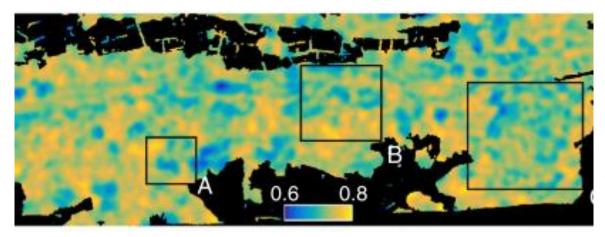
#### Coupling - Laserscanning (UAVLS) & Terrestrial Laser



#### Laegeren Forest, Switzerland

Morsdorf, F., Kükenbrink, D., Schneider, F.D., Abegg, M., Schaepman, M.E., 2018. Close-range laser scanning in forests: towards physically based semantics across scales. Interface Focus 8, 20170046. doi:10.1098/rsfs.2017.0046

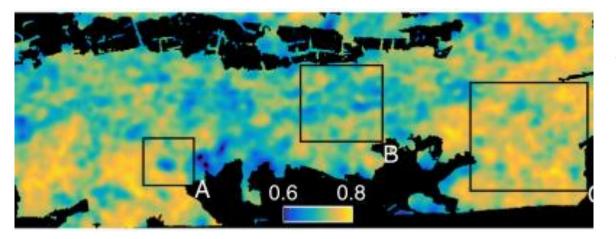
### **Ecosystem health by RS – Structural & Functional Diversity**



#### Morphological forest traits

Plant area index (PAI, blue), Canopy height (CH, red) Foliage height diversity (FHD, green)

#### →Morphological Evenness



#### **Physiological forest traits**

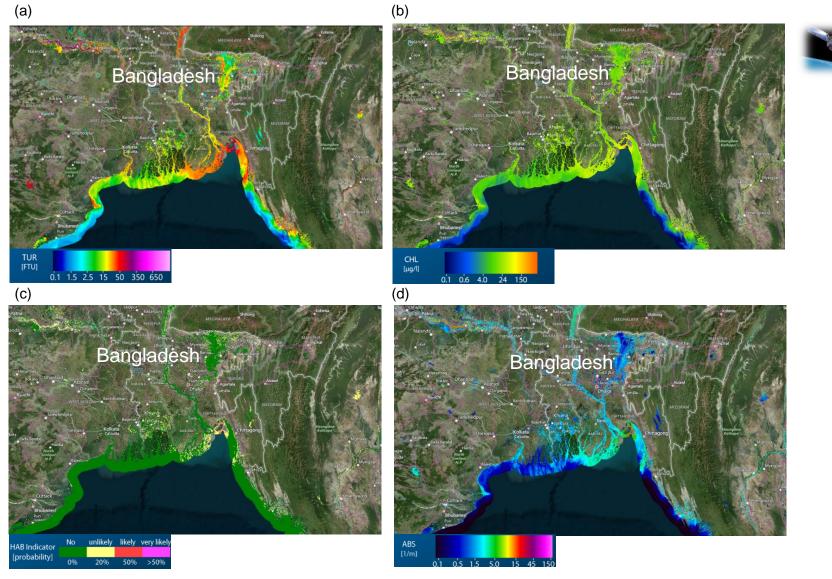
Water thickness (EWT, blue) Carotenoids (CAR, red) Chlorophyll (CHL, green)

#### → Physiological Evenness

Schneider, F.D., Morsdorf, F., Schmid, B., Petchey, O.L., Hueni, A., Schimel, D.S., Schaepman, M.E., 2017. Mapping functional diversity from remotely sensed morphological and physiological forest traits. Nat. Commun. doi:10.1038/s41467-017-01530-3

# Ecosystem health by RS – Water characteristics and

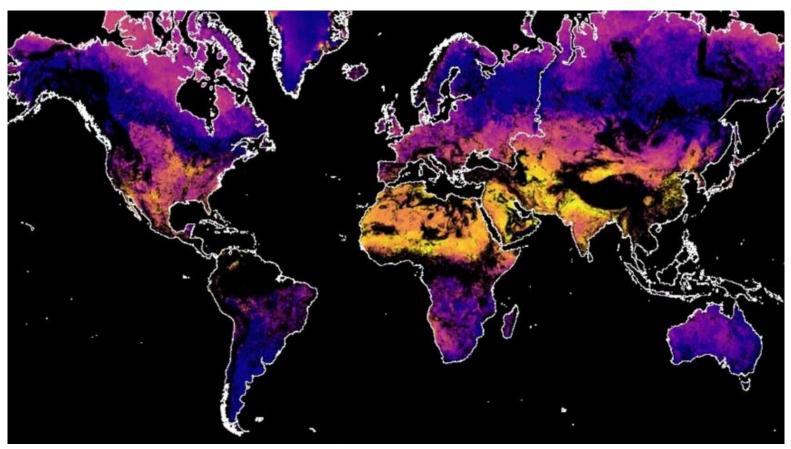
#### water quality indicator



http://www.worldwaterquality.org/

#### Earth Engine Data Catalog - Atmospheric methane (CH<sub>4</sub>) concentration

**Earth Engine's public data** archive **includes more than forty years of historical imagery** and scientific datasets, updated and expanded daily.



#### **Sentinel-5P Methane**

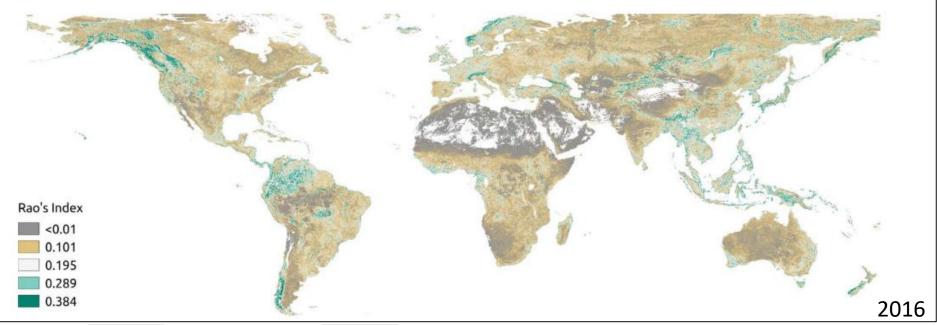
Atmospheric methane ( $CH_4$ ) concentration. After carbon dioxide ( $CO_2$ ), it is the most important contributor to the anthropogenically enhanced greenhouse effect. It enters Earth's atmosphere through both natural and anthropogenic processes, though the majority is of anthropogenic origin.

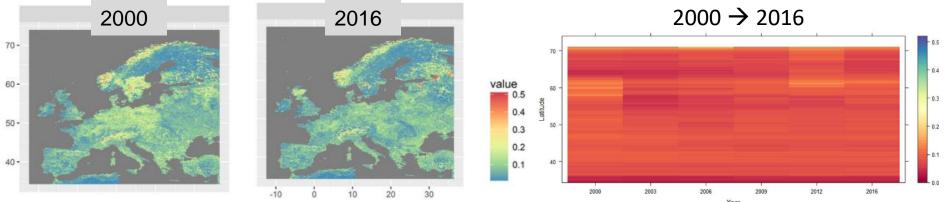
Dataset availability: 2019-02-08 – Present

https://developers.google.com/earth-engine/datasets

### **Ecosystem health by RS – Rao's quadratic diversity metric**

Rao's quadratic diversity metric - NDVI map of the world 2016 - http://land.copernicus.eu/global/products/ndvi



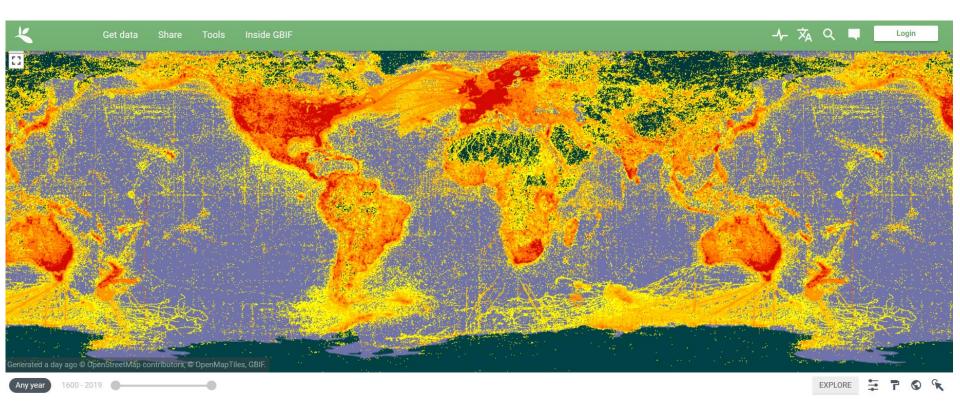


Rocchini, D., et al., 2018. Remotely sensed spatial heterogeneity as an exploratory tool for taxonomic and functional diversity study. Ecol. Indic. 85, 983–990. https://doi.org/10.1016/j.ecolind.2017.09.055

#### **GBIF**—the Global Biodiversity Information Facility

https://www.gbif.org/

—is an **international network** and research infrastructure funded by the world's governments and aimed at providing anyone, anywhere, **open access** to data about all types of life on Earth.



GBIF—the Global Biodiversity Information Facility

### And what is the problem – in Monitoring Bio- Geodiversity and ESS ?



Spaceborne



Airborne





UAV - Drone



Camera trap



Wireless-Sensor-Network (WSN)

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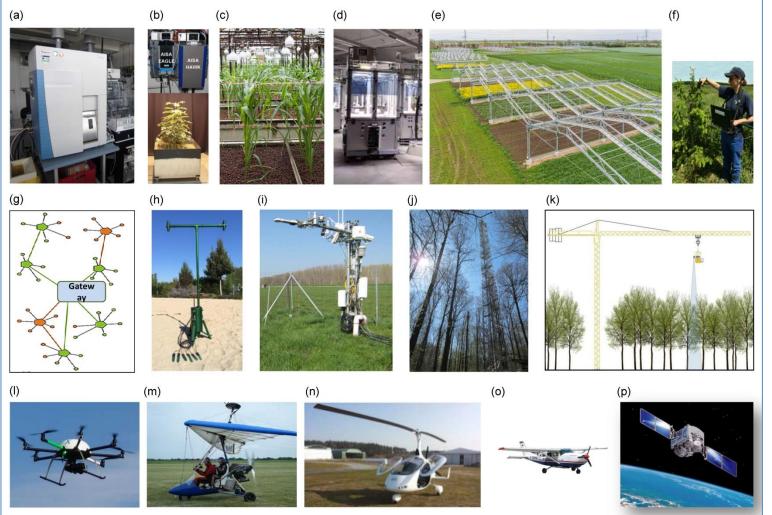
# **Data Science for Bio-Geodiversity - Challenges**

 Landscapes, Ecology, Bio-and Geodiversity, Processes & Functions are: Complex, multidimensional, multi-scale and mostly non-linear

- Not one monitoring approach, monitoring platform, model, space-time scale, tool or data alone is sufficient to explain the complexity of landscapes, processes or functions
- We have to look for necessary requirements dealing of Complexity, Multidimensionality …,



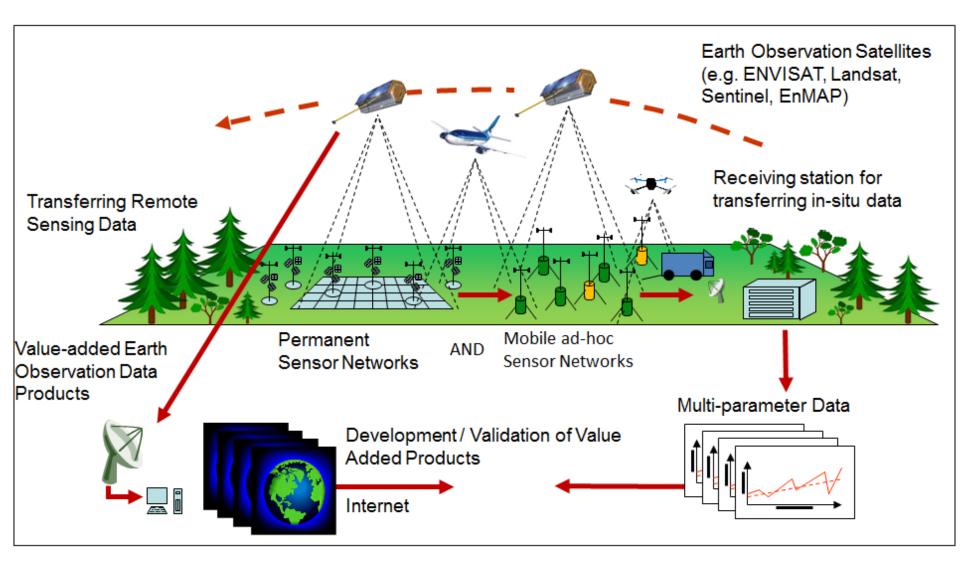
### **Data Science – Requirement – Remote Sensing**



use & couple different plattforms

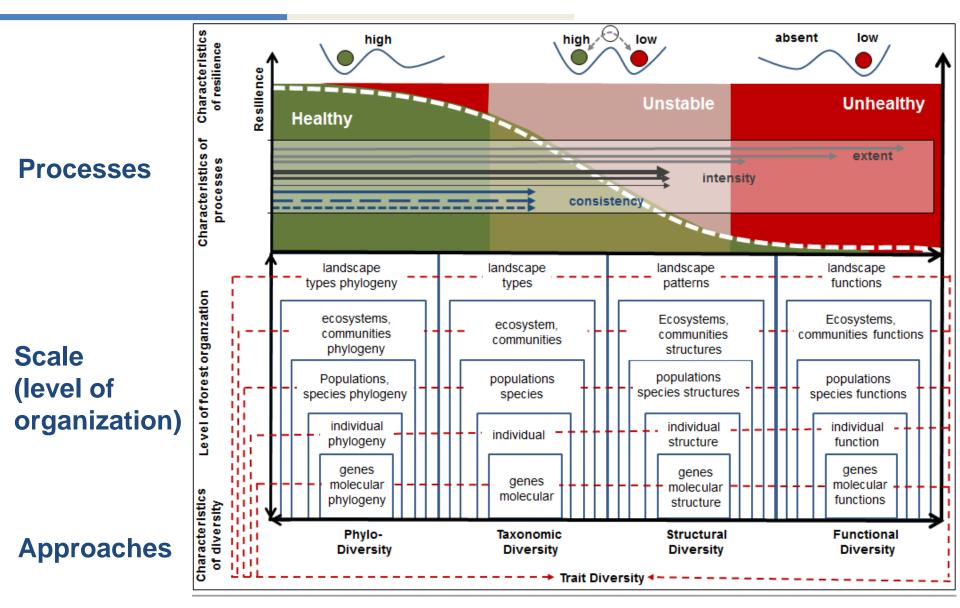
**Lausch** et al., 2020. A range of Earth Observation techniques for assessing plant diversity Jeannine Cavender-Bares, John Gamon, Philip Townsend (eds): The nature of biodiversity: prospects for remote detection of genetic, phylogenetic, functional and ecosystem components and importance in managing Planet, Jeannine Cavender-Bares, John Gamon, Philip Townsend, Springer (in press)

### **Data Science – Requirement – Coupling RS Platforms**

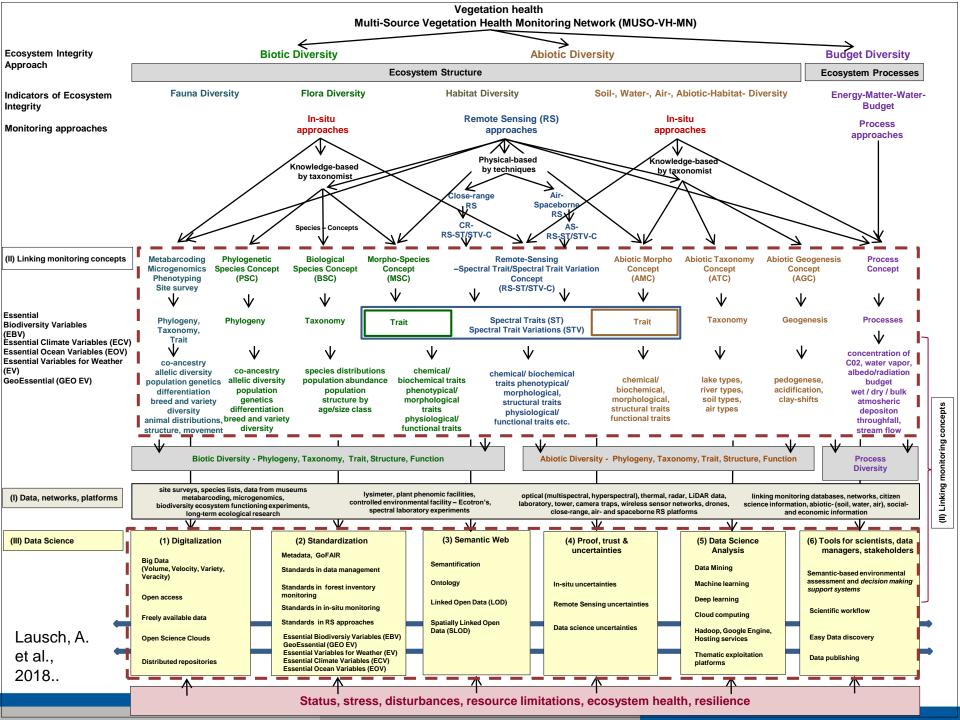


Lausch, A. et al., 2018. Understanding Forest Health with Remote Sensing, Part III: Requirements for a Scalable Multi-Source Forest Health Monitoring Network Based on Data Science Approaches. Remote Sensing, 10, 1120

#### Approaches, scale, processes $\rightarrow$ Biodiversity and their resillience



Lausch, A. et al., 2018. Understanding Forest Health with Remote Sensing, Part III: Requirements for a Scalable Multi-Source Forest Health Monitoring Network Based on Data Science Approaches. Remote Sensing, 10, 1120; doi:10.3390/rs10071120.



### **Data Science – Challenge - Digitalization**

### **Ecology and Evolution**

Open Access

BIOLOGICAL

### The PREDICTS database: a global database of how local terrestrial biodiversity responds to human impacts

Lawrence N. Hudson<sup>1</sup>\*, Tim Newbold<sup>2,3</sup>\*, Sara Contu<sup>1</sup>, Samantha L. L. Hill<sup>1,2</sup>, Igor Lysenko<sup>4</sup>, Adriana De Palma<sup>1,4</sup>, Helen R. P. Phillips<sup>1,4</sup>, Rebecca A. Senior<sup>2</sup>, Dominic J. Bennett<sup>4</sup>, Hollie Booth<sup>2,5</sup>, Argyrios

#### Global Change Biology

Global Change Biology (2011) 17, 2905–2935, doi: 10.1111/j.1365-2486.2011.02451.x

#### TRY – a global database of plant traits

J. KATTGE\*, S. DÍAZ†, S. LAVOREL‡, I. C. PRENTICE§, P. LEADLEY¶, G. BÖNISCH\*, F. GARNIER‼ M. WESTORYS, P. R. REICH\*\*, ‡†, I. I. WRIGHTS, I. H. C. CORNELISSEN!!!

#### World database of protected areas (WDPA)

Movebank – For Animal Tracking data www.movebank.org

#### Encyclopedia of life (EOL)



Contents lists available at ScienceDirect

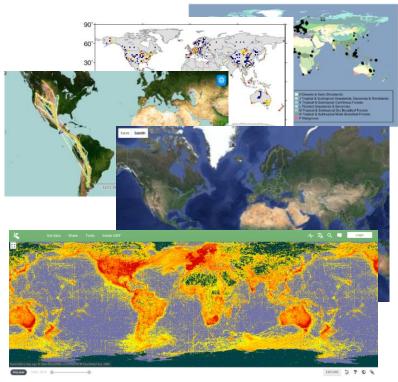
**Biological Conservation** 

journal homepage: www.elsevier.com/locate/biocon

Perspective

Free and open-access satellite data are key to biodiversity conservation

W. Turner<sup>a,\*</sup>, C. Rondinini<sup>b</sup>, N. Pettorelli<sup>c</sup>, B. Mora<sup>d</sup>, A.K. Leidner<sup>a,e</sup>, Z. Szantoi<sup>f</sup>, G. Buchanan<sup>g</sup>,



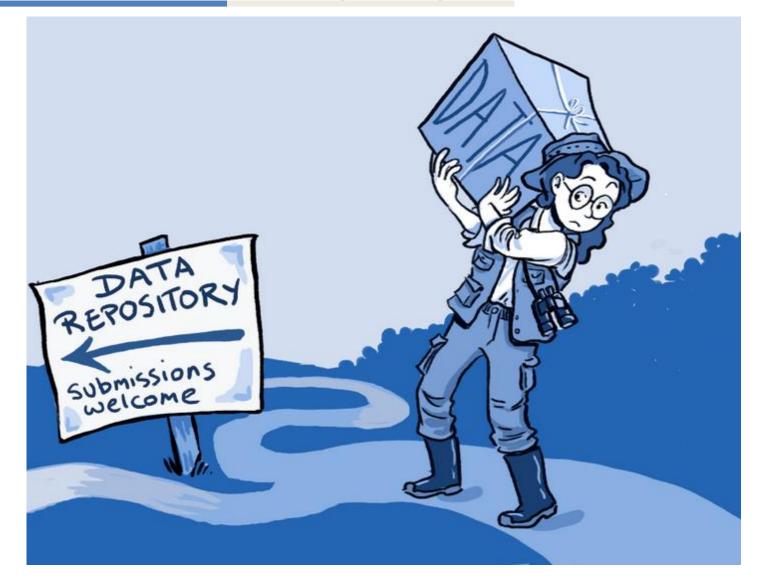
**Big Data** 

Free Data

**Open Data** 

**Complex Data** 

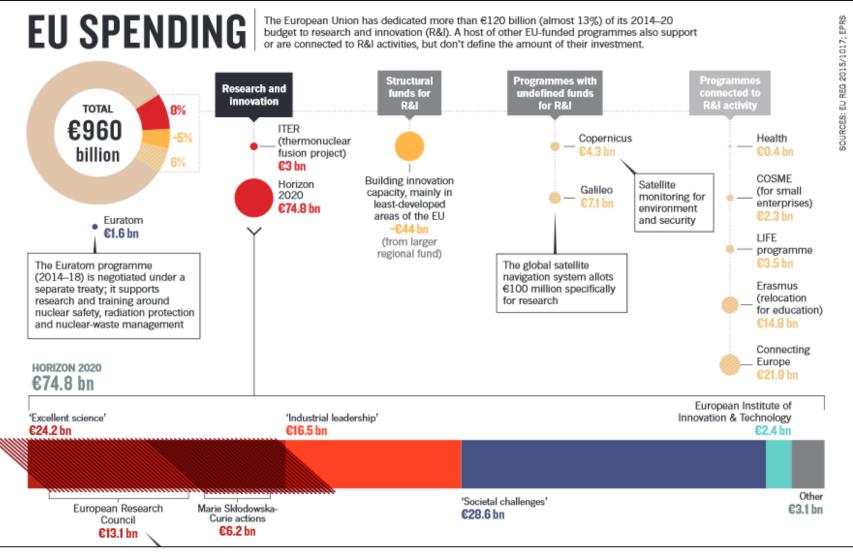
**Data Science – Challenge - Digitalization** 



Roche, D.G., Lanfear, R., Binning, S.A., Haff, T.M., Schwanz, L.E., Cain, K.E., Kokko, H., Jennions, M.D., Kruuk, L.E.B., 2014. Troubleshooting Public Data Archiving: Suggestions to Increase Participation. PLoS Biol. 12. doi:10.1371/journal.pbio.1001779

### **Data Science – Challenge - Digitalization**

### Investitions/EU - Data-Generation 2014-2020 → 120B €



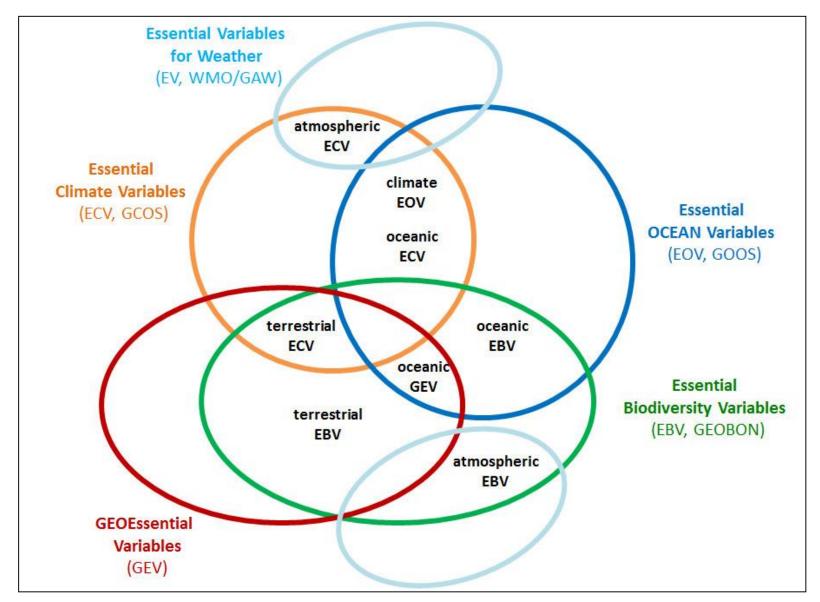
Abbott, A.; Butler, D.; Gibney, E.; Schiermeier, Q.; Van Noorden, R. Boon or burden: done for science? Nature 2016, 534, 307–309. Although 90% of the world's data was generated over two years, around

→ "50% of all research and experiment data (= US\$28B/year) are never found again !

→ and over 80% of it never makes it to a trusted and sustainable repository" (Ayris et al., 2016)

Ayris, P.; Berthou, J.-Y.; Bruce, R.; Lindstaedt, S.; Monreale, A.; Mons, B.; M. Tochtermann, K.; Wilkinson, R. Realising the European Open Science Cloud; Europe Brussels, Belgium, 2016.

#### **Data Science – Standardization in Monitoring**



Lausch, A. et al., 2018. Understanding Forest Health with Remote Sensing, Part III: Requirements for a Scalable Multi-Source Forest Health Monitoring Network Based on Data Science Approaches. Remote Sensing, 10, 1120

## **Data Science – Requirement – Metadata/Data - FAIR**

#### www.nature.com/scientificdata

# SCIENTIFIC DATA

SUBJECT CATEGORIES

#### » Research data » Publication

#### OPEN Comment: The FAIR Guiding Principles for scientific data management and stewardship characteristics

Mark D. Wilkinson et al."

Received: 10 December 2015 Accepted: 12 February 2016 Published: 15 March 2016

There is an urgent need to improve the infrastructure supporting the reuse of scholarly data. A diverse set of stakeholders-representing academia, industry, funding agencies, and scholarly publishers-have come together to design and jointly endorse a concise and measureable set of principles that we refer to as the FAIR Data Principles. The intent is that these may act as a guideline for those wishing to enhance the reusability of their data holdings. Distinct from peer initiatives that focus on the human scholar, the FAIR Principles put specific emphasis on enhancing the ability of machines to automatically find and use the data, in addition to supporting its reuse by individuals. This Comment is the first formal publication of the FAIR Principles, and includes the rationale behind them, and some exemplar implementations in the community.

### Findable

Accessible

### Interoperable

#### Reusable

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#### Box 2 | The FAIR Guiding Principles

#### To be Findable:

- F1. (meta)data are assigned a globally unique and persistent identifier
- F2. data are described with rich metadata (defined by R1 below)
- F3. metadata clearly and explicitly include the identifier of the data it d
- F4. (meta)data are registered or indexed in a searchable resource

#### To be Accessible:

- A1. (meta)data are retrievable by their identifier using a standardized co
- A1.1 the protocol is open, free, and universally implementable
- A1.2 the protocol allows for an authentication and authorization procedure, where necessary
- A2. metadata are accessible, even when the data are no longer available

#### To be Interoperable:

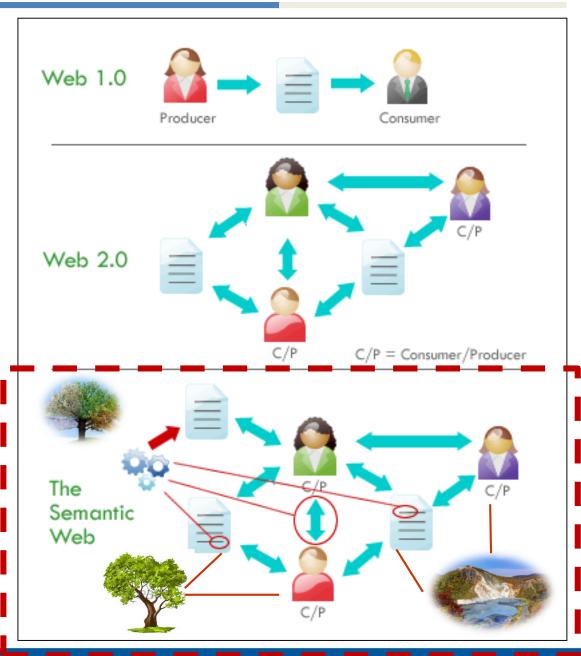
- 11. (meta)data use a formal, accessible, shared, and broadly applicable language for knowledge representation.
- 12. (meta)data use vocabularies that follow FAIR principles
- (meta)data include gualified references to other (meta)data

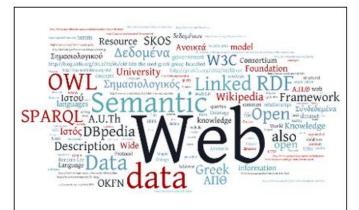
#### To be Reusable:

- R1. meta(data) are richly described with a plurality of accurate and relevant attributes
- R1.1. (meta)data are released with a clear and accessible data usage license
- R1.2. (meta)data are associated with detailed provenance
- R1.3. (meta)data meet domain-relevant community standards

Wilkinson, M.D., Dumontier, M., Aalversberg, I.J., Appleton, G., Axton, M., 2016. Comment : The FAIR Guiding Principles for scientific data management and stewardship. Nat. Commun. 3:160018.1-9.

## **Data Science – Requirement - Semantification**





# Semantic Web / Linked Open Data

Handling: ≻ Complex-Data



# Data Science – Requirements – Information Management



Environmental Research Infrastructures Providing Shared Solutions for Science and Society

UFZ

http://www.envriplus.eu/

Scientific Discover problem data Analysis publication Need to support user centered research activities Workflow Execution design and debug Data Need to manage data in its lifecycle acquisition Data Data use curation Data Data processing publishing Need to manage infrastructure Request resources, e.g., computing, storage Support and networks Monitoring Scheduling

Zhiming Zhao, 2018,

# **Data Science – Requirements**

#### Coupling to the Research infrastructure in enviornmental and earth sciences



Environmental Research Infrastructures Providing Shared Solutions for Science and Society

HELMHOLTZ ZENTRUM FÜR UMWELTFORSCHUNG UFZ

Zhiming Zhao, 2018,

### Data Science – Requirements – for ESS

- **Good Indicators** for ESS, environmental changes, stress & disturbances
- Coupling in-situ and RS approaches
- Digitalization

(Big Data (Volume, Velocity, Variety, Veracity), Open Access, Freely available data, Open Science Clouds, Distributed repositories, TEP – Thematic Exploitation Platform – ESA)

**Comparable Data and Data bases** 

(Google Engine, GBIF—the Global Biodiversity Info Copernicus RS Data, DEM .....)

Standardization

(Metadata, GoFAIR, Concept of Essential Variables

Semantification

(Semantic Web/Web 4.0, Ontology; Linked Open Da

Data Science Analysis

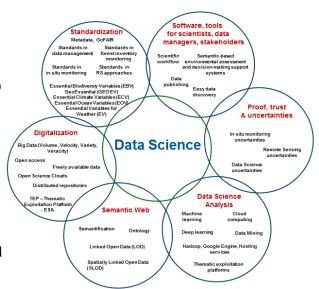
(Machine Learning, Deep learning, Cloud Computing,

Data Mining, Hadoop, Google Engine, Hosting services)

Proof, trust & uncertainties

(In-situ monitoring, Remote Sensing & Data Science uncertainties)

• Easy software, tools for data manager, stakolders & for politics



Lausch, A. et al., 2018.. Remote Sensing

#### Bio- Geodiversity, Ecosystem health, RS, Data Science - Paper

Lausch, A., Schmidt, A., Tischendorf, L., 2015. Data mining and linked open data – New perspectives for data analysis in environmental research. Ecol. Modell. 295, 5–17. https://doi.org/10.1016/j.ecolmodel.2014.09.018

Lausch, A. et al., 2016. Linking Earth Observation and taxonomic, structural and functional biodiversity: Local to ecosystem perspectives. Ecol. Indic. doi:10.1016/j.ecolind.2016.06.022

Lausch, A., et al., 2016. Understanding Forest Health with Remote Sensing -Part I—A Review of Spectral Traits, Processes and Remote-Sensing Characteristics. Remote Sens. 2016, Vol. 8, Page 1029 8, 1029. doi:10.3390/RS8121029

Lausch, A., et al., 2017. Understanding Forest Health with Remote Sensing-Part II—A Review of Approaches and Data Models. Remote Sens. 9, 129. doi:10.3390/rs9020129

**Lausch, A.** et al., **2018**. Understanding Forest Health with Remote Sensing, Part III: Requirements for a Scalable Multi-Source Forest Health Monitoring Network Based on Data Science Approaches. Remote Sensing, 10, 1120; doi:10.3390/rs10071120.

**Lausch, A.** et al., **2018**. Understanding and assessing vegetation health by in-situ species and remote sensing approaches. Methods in Ecology and Evolution, 00: 1–11. doi.org/10.1111/2041-210X.13025.

Lausch, A. et al., 2019. Linking Remote Sensing and Geodiversity and Their Traits Relevant to Biodiversity— Part I: Soil Characteristics. Remote Sens. 11, 2356. https://doi.org/10.3390/rs11202356



#### Monitoring Bio- Geodiversity and Ecosystem health by Traits, Remote Sensing (RS) and Data Science approaches



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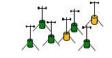
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