

# Most of the world's bioimaging data lacks a clear path to being shared.

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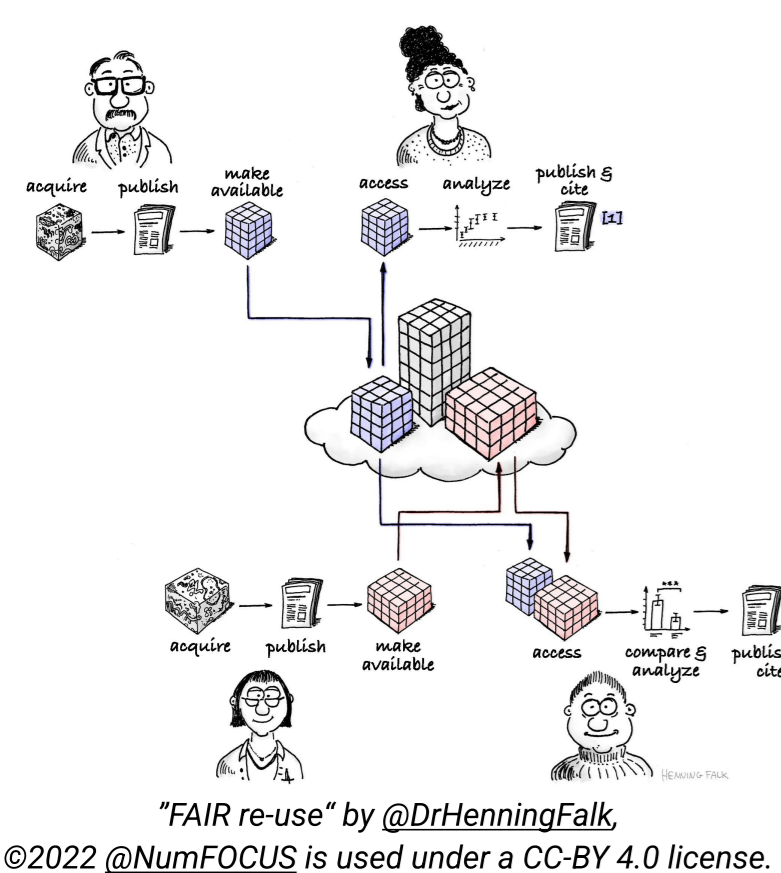
German BioImaging e.V. Society for Microscopy and Image Analysis, Konstanz, Germany

0000-0003-4028-811X

*AI's Dirty Little Secret: Without FAIR Data, It's Just Fancy Math*

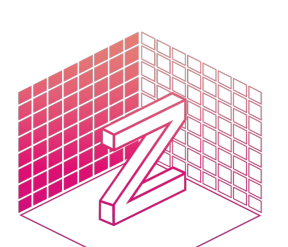
**Restricted access** or a **limited** number of biological images and their metadata inhibits researchers from developing robust and generalizable models, potentially **decreasing** the **accuracy** and **performance** of AI applications in bioimaging.

**Community Repositories** are available to **share** and **find valuable data**. Make use of them! **1 2 3 4**



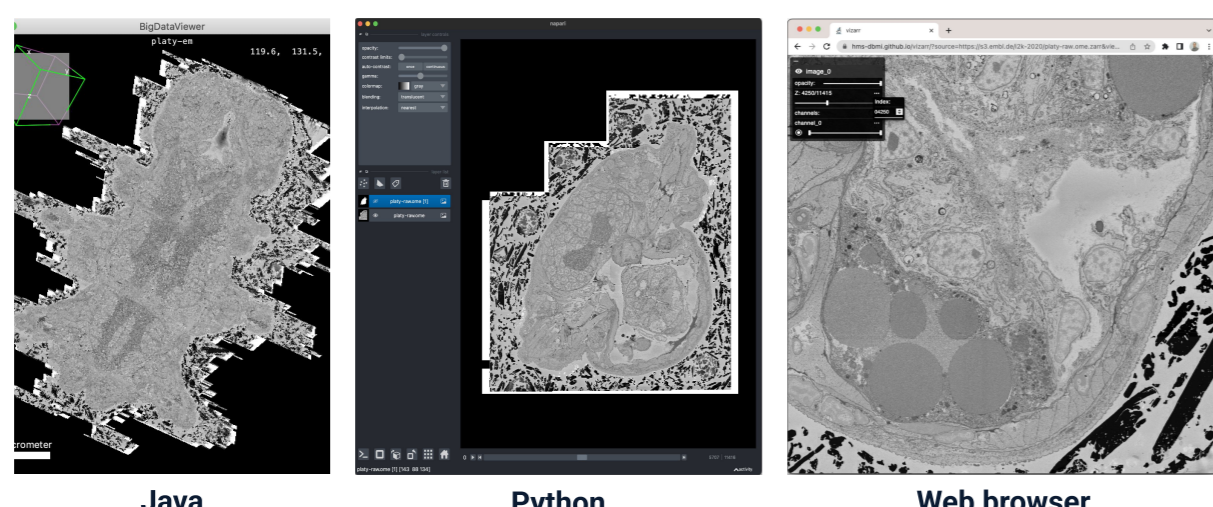
**FAIR Image Objects:** FDO-compatible datatype for bioimaging to ensure data is **open** and **web-accessible** to enable **efficient distributed processing**. **1 4 5 6**

**Terabytes** of pixels as well as analytical results can be made shareable, linkable, browsable, re-usable, archivable. A **pyramidal** structure allows Google Maps-style zooming. A cloud-optimized ("chunked") format allows referencing individual regions of an image in parallel.



<https://zarr.dev>

"larger-than-memory, n-dimensional, typed arrays"



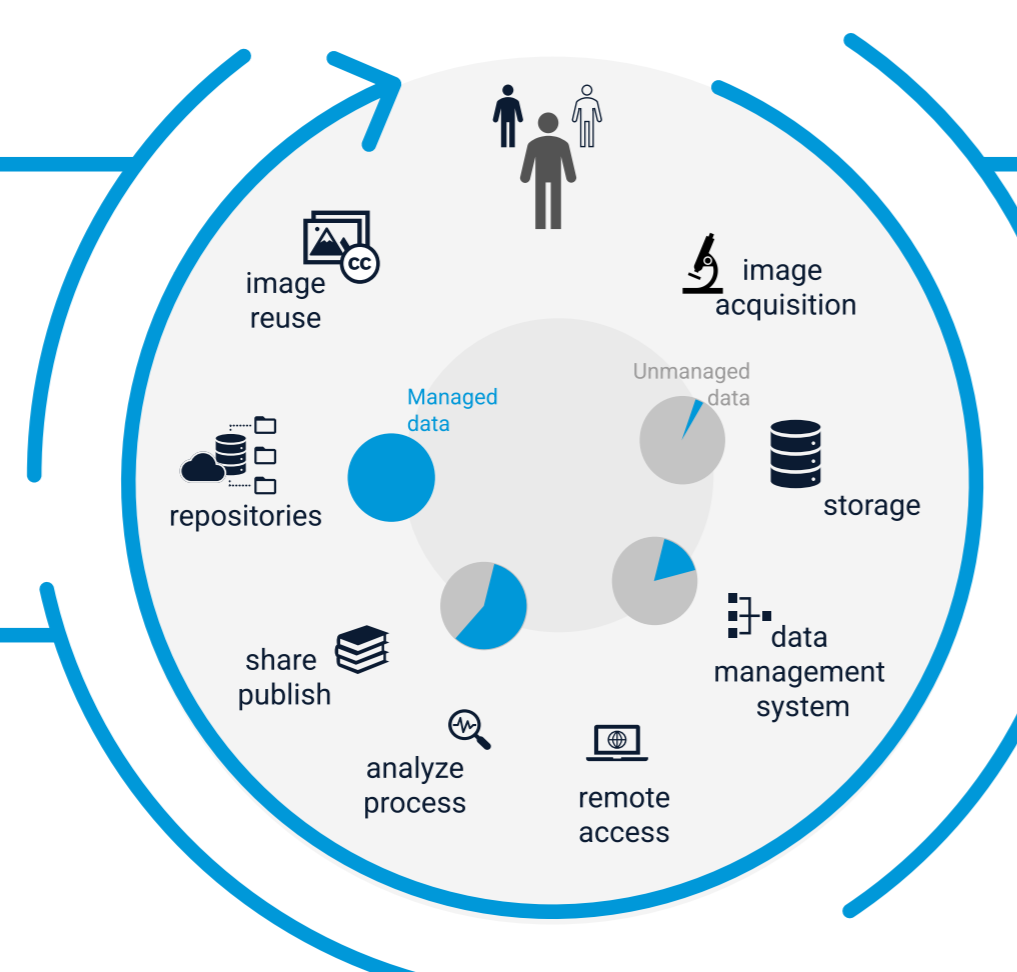
8-TB electron microscopy volume of a 6 day old *Platymeris* larva from Vergara et al. 2020 available at: <https://s3.embl.de/f2k-2020/platy-raw.ome.zarr>

See QR code for demo: <https://wklink.org/6422>

**NFDI4BIOIMAGE** is one example of a national endeavor to **organize existing data** for use in AI. **1 5 6 7**

**Objective 4**  
Capacitate researchers for FAIR image data management

**Objective 3**  
Maximize the reach of reproducible image analysis workflows in the community



**Objective 1**  
Champion the **standardization** of the „bioimage data“ type

**Objective 2**  
Provide scalable **infrastructure** for FAIR image data

## ...Data Requirements for AI

### Accessibility<sup>1</sup>

- Foster innovation
- Model improvement

### Quality<sup>2</sup>

- AI model accuracy
- Scaling of AI applications

### Variety<sup>3</sup>

- Improved generalization
- Enhanced adaptability
- Mitigation of biases
- Facilitation of cross-domain adoption

### Context<sup>4</sup>

- Data interpretation
- Ability to learn
- Transparent and comprehensible decisions
- Better handling of unexpected or varying conditions

### Structure<sup>5</sup>

- Efficient processing of large amounts of information
- Reusability of data and thus interoperability of AI applications

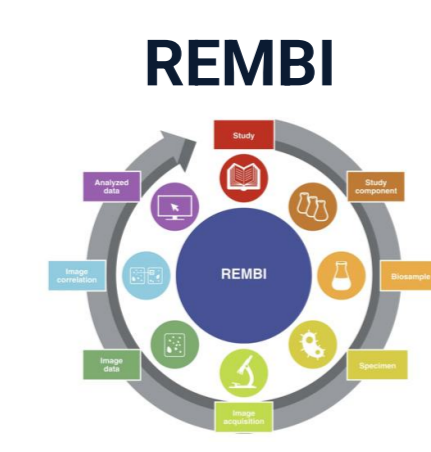
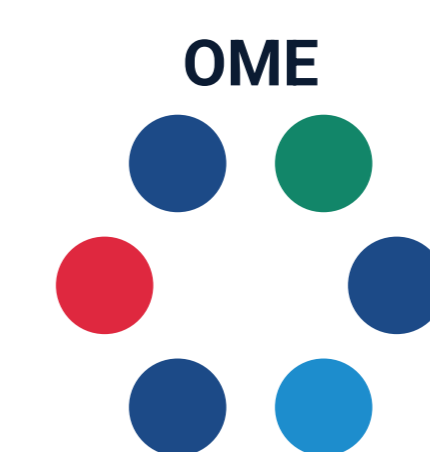
### Formats<sup>6</sup>

- Scalability
- Compatibility
- Storage space optimization
- Efficient processing

### Infrastructure<sup>7</sup>

There are rarely comprehensive **metadata** associated with image data and a **lack of semantic data integration** means AI can't assign meaning to bioimages.

**Community metadata standards** and techniques are available to **facilitate and automate** semantic data integration. **1 2 3 4**



**Semantic data integration** enables the **integration and harmonization of data** from various bioimaging sources and related data. **1 4 5 6**

This approach focuses on understanding the **relationships** and **meaning** of the data elements to facilitate efficient data sharing, analysis, and interpretation in the field of bioimaging.

For successful semantic data integration, certain requirements for metadata are essential:

- Unique identifiers (PID)
- Relationships and Linkages (RDF)
- Semantic annotations (Ontologies)
- Metadata standards (schemas, models)

Such metadata enables **mapping** of different **schemas through ontologies** and realizes **entity mapping** based on a **knowledge graph**.

If this metadata is missing, this requires time-consuming subsequent labeling to create the context required for the AI.



"How many humans does it take to make tech seem human? **Millions**"

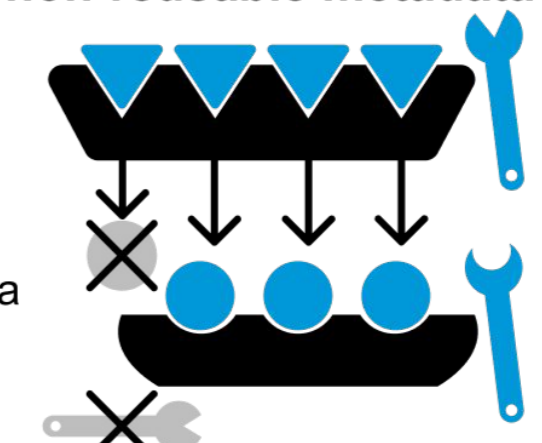
Engagement with **hardware vendors** is needed to enable **complete and re-usable metadata** directly from acquisition systems. **1 2 4**

### Current Situation: non-reusable metadata

Proprietary (e.g., **vendor**) metadata are restricted to vendors tools.

**Community metadata** cannot represent all vendor-specific information.

- Legend**
- Core models: provide general building blocks that can be re-used
  - Community models: community approved metadata models that enable open-source tools.
  - ▲ Custom models: extensions that are in development or highly specialized in their application.



### Next-Generation Metadata Framework

With a **modular** framework, all metadata can be recorded in a common framework.

Metadata is **accessible** by community and vendor tools.

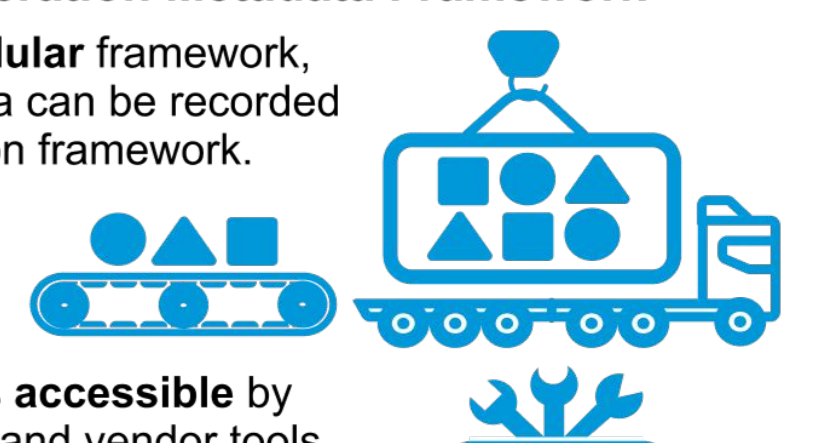


Figure is co-authored with Caterina Strambio De Castillia, CC-BY