

# From Pixels to Policy: Research Data Management Strategies for AI-Driven Bacterial Detection in Food Safety Research

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BRICCs-RDM Conference 2025



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# The Challenge: Food Safety Meets AI

Traditional bacterial detection: **24-72 hours** to produce results

AI-driven approach: Detection in **minutes** using YOLOv8 architecture

Target pathogens: **E. coli** and **Salmonella** in food samples

The data challenge: **Massive image datasets (2TB+)** with complex metadata



# Our Use Case: YOLOv8 Bacterial Detection System

**High-resolution microscopic imaging** with standardized Gram staining procedures

Mixed cultures with **temporal data (0.5-4 hours)** to capture bacterial growth dynamics

Complex sample preparations ("**with onion**" and "**without onion**") to reflect real-world scenarios

Individual images: **50-150 MB each**, cumulative dataset exceeding **2 TB**

Manual annotation: **30-60 minutes per image** requiring expert knowledge



# Processing Challenges in AI-Intensive Research

**Scale:** 2TB+ datasets with individual files exceeding 100MB, overwhelming conventional research infrastructure

**Solution:** Distributed processing pipeline with parallel image processing nodes

**Complexity:** Complicated metadata interdependencies capturing sample preparation, imaging parameters, and experimental context

**Solution:** GPU-accelerated image processing reducing time from hours to minutes

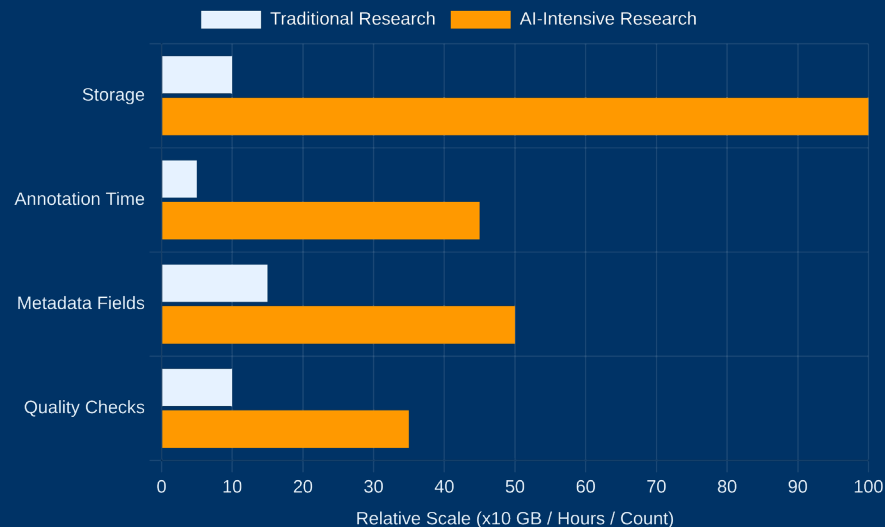
**Quality:** Balancing biological variability with standardization needs for AI model training

**Solution:** Automated quality metrics with threshold-based filtering

**Version control:** Traditional systems inadequate for massive binary files, requiring custom workflows

**Solution:** Standardized processing workflows with version control

AI Research vs Traditional Research Requirements



# FAIR Implementation Strategies

**Findability:** Controlled vocabularies, persistent identifiers (DOIs), domain-specific search interfaces

**Accessibility:** Tiered access control systems, data use agreements, long-term preservation services

**Interoperability:** Standardized formats (TIFF, PNG), metadata encoding (Dublin Core, DataCite), COCO extensions

**Reusability:** Comprehensive provenance documentation (PROV-O), Creative Commons licensing, analytical tools



# Data Processing Tools & Workflows

## Image Processing Tools

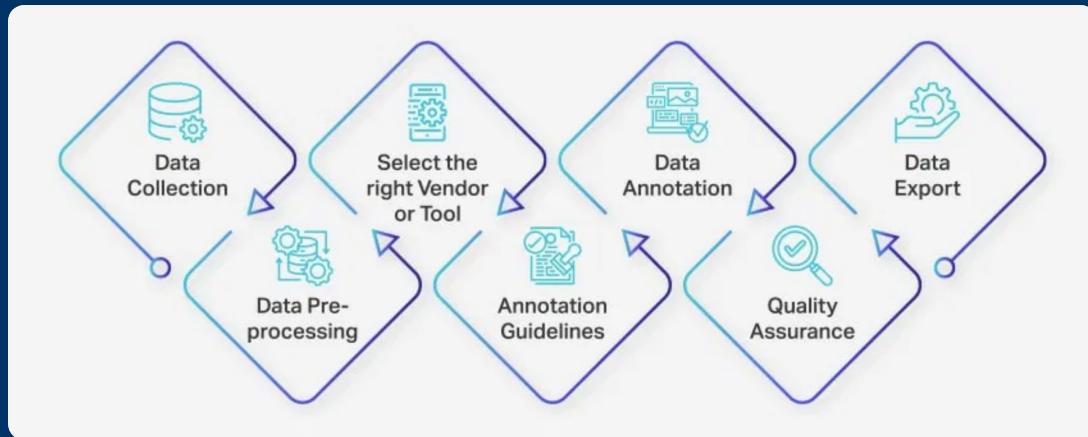
- ✓ **OpenCV:** Core library for image preprocessing and enhancement
- ✓ **scikit-image:** Quality assessment and feature extraction

## Annotation Management

- ✓ **LabelImg:** Custom-modified for bacterial annotation
- ✓ **CVAT:** Collaborative verification platform

## Version Control Systems

- ✓ **DVC (Data Version Control):** For large binary files
- ✓ **Git LFS:** For metadata and configuration



# Lessons

**Scale Matters:** Traditional data processing tools fail with AI-scale datasets - invest in scalable infrastructure from the start

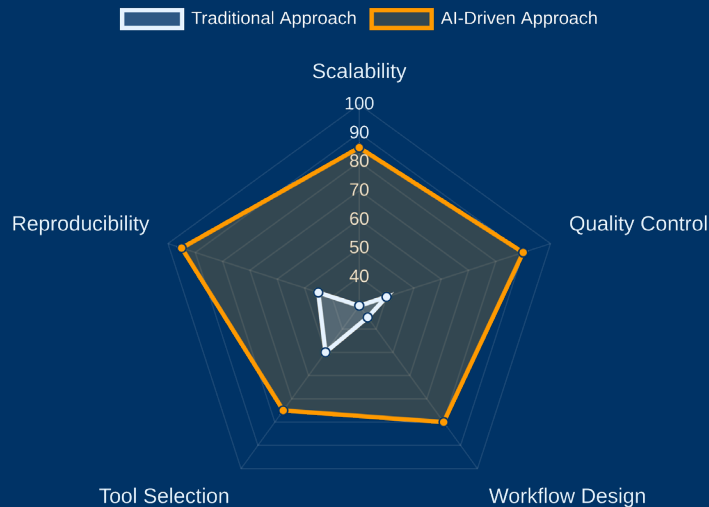
**Quality First:** Automated quality control is essential - poor quality data compounds errors throughout the pipeline

**Workflow Design:** Systematic processing pipelines prevent errors and save time - document every step and automate where possible

**Tool Selection:** Right tools make the difference between success and failure - evaluate tools based on your specific data characteristics

**Reproducibility:** Good processing documentation enables scientific reproducibility - version control everything, including processing parameters

Traditional vs. AI-Driven Data Processing Approaches



# Requirements & Recommendations

## Standardized Metadata Templates

Machine-readable schemas with automated validation for AI-intensive research

## Comprehensive RDM Policies

Address data collection, storage, processing, sharing, and preservation throughout research lifecycles

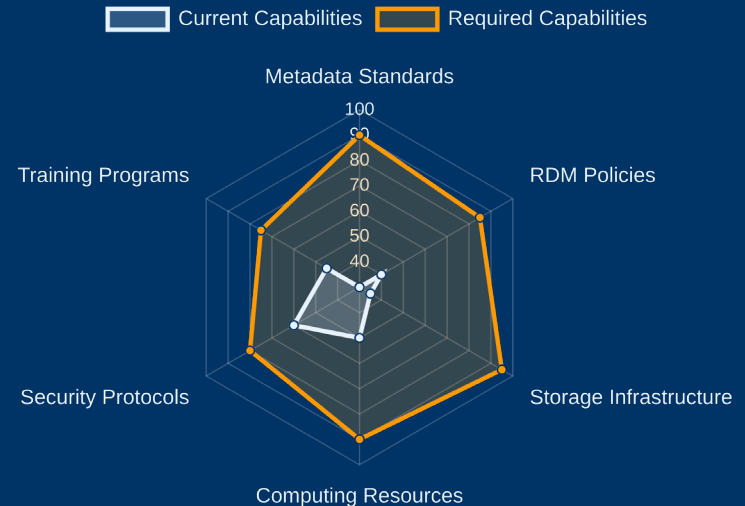
## Cyberinfrastructure Development

GPU-accelerated computing, specialized storage systems, and annotation management platforms

## Security Infrastructure

Protect sensitive research data while maintaining accessibility for collaborative research

### Gap Analysis: Current vs. Required Capabilities



# Conclusion & Call to Action

RDM is an **enabler of scientific progress**, not merely an administrative requirement

The stakes extend beyond academia to **public health, food security, and economic stability**

The **window of opportunity is limited** as AI technologies advance rapidly

**Sustained commitment** needed from research institutions, funding agencies, and the scientific community



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## Building the Future

Develop standardized metadata templates

Establish comprehensive RDM policies

Invest in specialized cyberinfrastructure

Collaborate on international standards



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