Facility and Instrument PIDs in the Materials Research Data Domain

Is there a roadmap? Are we optimistic or pessimistic?

David Elbert: elbert@jhu.edu

CDO PARADIM Materials Innovation Platform (MIP) ARL HTMDEC Extreme Data Pl 2 NSF DMREF Co-Pl DOE Supported Catalysis Co-Pl Materials Research Data Alliance (MaRDA) MaRCN FAIROS-RCN Pl



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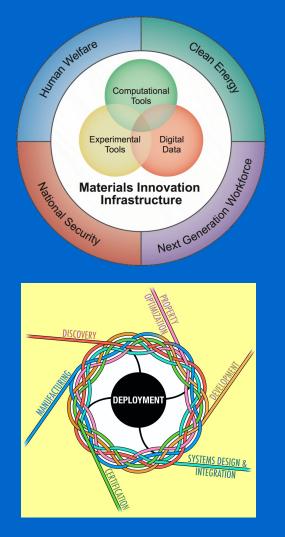


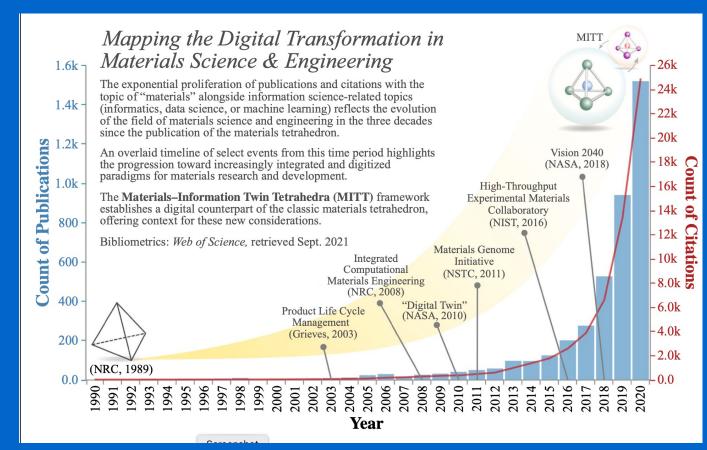
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Got PID?

Materials Genome Initiative (MGI) Driven Directions





Deagen et al, 2022 MRS Bulletin doi:10.1557/s43577-021-00214-0







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Materials Science Motivations

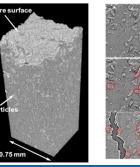
Firehose of Data

- Higher resolution
- Shorter time scales
- Higher dimensionality
- Dynamic experiments
- Larger simulations
- Tighter processing control
- **Diverse, Distributed Data**
- Experimental
 - labs great and small
- Modeling
 - finite element to ab initio
 - desktops to supercomputers





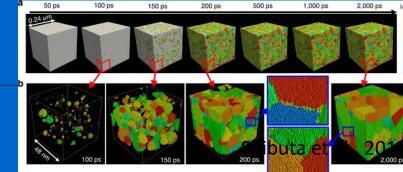






De Carlo et al., 2012





Materials Research Data Alliance

https://marda-alliance.org

MaRCN

MGI 2.0: People are infrastructure, too! 2021 Revised Strategic Plan Goal 1, Objective 2:

Unify the Materials Innovation Infrastructure
Bridge, Build, and Bolster Elements of the MII.
Establish a National Materials Data Network
Accelerate through National Grand Challenges







MaRDA is a *community network* focused on developing the *open, accessible, and interoperable materials data* that fuels the Materials Genome Initiative (MGI). MaRDA is a **convergence of people and ideas working together** to connect materials data infrastructure to accelerate discovery, enable new insights into materials mechanisms, and lay a foundation for both human-centered and artificial intelligence-assisted approaches to materials design.







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FAIR for Materials:

FAIR is still poorly understood by most of our investigators
They don't know that FAIR are principles not implementation



Community action on FAIR data will fuel a revolution in materials research

L. Catherine Brinson,*[©] Laura M. Bartolo, Ben Blaiszik, David Elbert, Ian Foster, Alejandro Strachan, and Peter W. Voorhees

Check for updates

Impact Opinion & Perspective

The FAIR Guiding Principles

Findable:

- F1 Data and metadata 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 describes F4 Data and metadata are registered or indexed in a searchable resource

Accessible:

- A1 Data and metadata are retrievable by their identifier using a standardized communications protocol
- 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 availabl

Interoperable:

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

Reusable:

- R1 Data and metadata are richly described with a plurality of accurate and relevant attributes
- R1.1 Data and metdata are released with a clear and accessible data usage license

Adapted from Wilkinson et al., 2016

https://fairtoolkit.pistoiaalliance.org/fair-guiding-principles/



https://doi.org/10.1557/s43577-023-00498-4







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FAIR for Materials

fAIR means metadata, too

Community action will fuel a revolution research

L. Catherine Brinson,*[®] Laura Ian Foster, Alejandro Strachan

https://doi.org/10.1557/s43577-023-00498-4





Example: Coordinated Development for Interoperability

Metadata Extractors

Matthew Evans, UC Louvain Peter Kraus, TU Berlin David Elbert, Johns Hopkins

Interoperability Layer

•LinkML

•Translate LinkML instance data to OWL (TBoxes and ABoxes)

github.com/marda-alliance/metadata_extractors									Q	Û	
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		.github			Add CODEOWNERS					3 months	ago
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MaRDA Extractors WG

This repository contains organizational info for a MaRDA working group (WG) focused on connecting and advancing interoperability of efforts on automated extraction of metadata from materials files.

Contacts:

- Matthew Evans, UCLouvain (matthew.evans[at]uclouvain.be)
- Peter Kraus, TU Berlin (peter.kraus[at]ceramics.tu-berlin.de)
- David Elbert, Johns Hopkins University (elbert[at]jhu.edu)

Contributing

This working group is completely open. If you would like to be added ot the mailing list please reach out to us over email, or just turn up at a meeting!

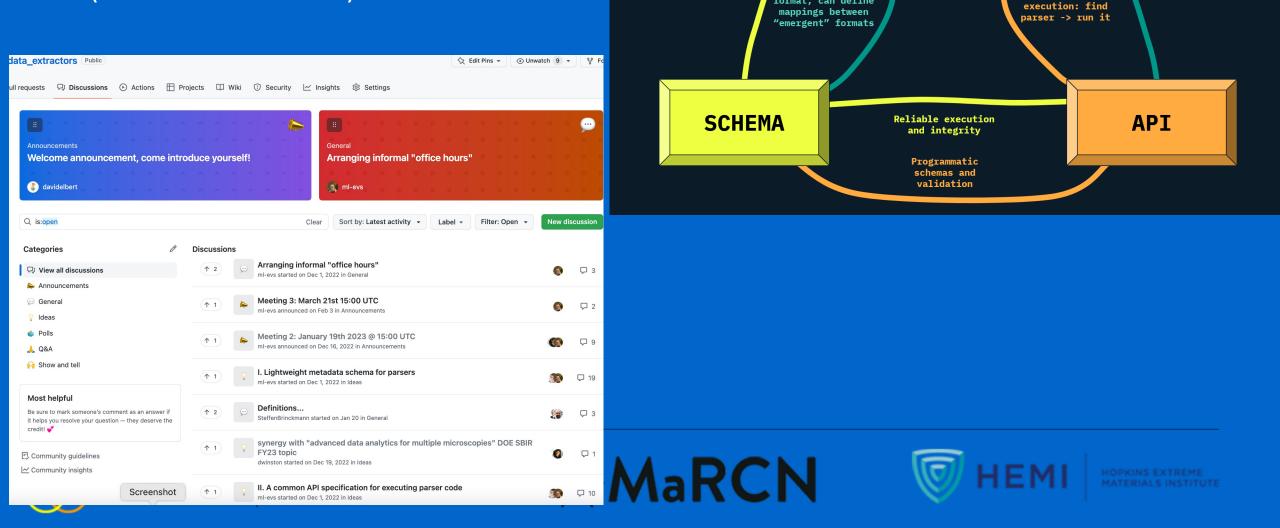
The GitHub discussions on this repo can be used for pretty much any related chat, and specific code suggestions/feedback can be made as pull requests to the GitHub repos for each subproject:

- marda-alliance/metadata_extractors_schema
- marda-alliance/metadata_extractors_registry
- marda-alliance/meta Screenshot pi

Interoperability Layer

•LinkML

•Translate LinkML instance data to OWL (TBoxes and ABoxes)



Ability to search/index over schemas and point to persistent (to

some degree) URL

Schema defines a

format, can define

REGISTRY

"Semi-centralized" conda-forge style recipes with tests

and auditing

Machine-actionable

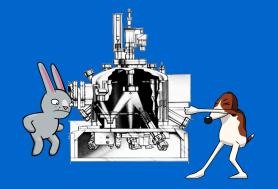
Scientists can be myopic

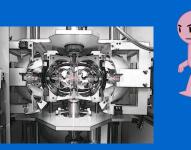
Many Moving Parts

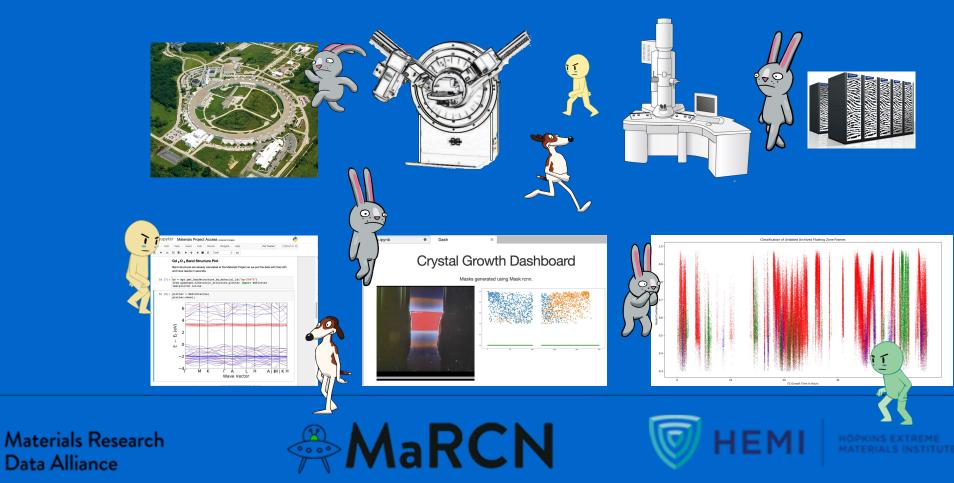
- Equipment
- People

aRI

• Ideas







Somas "Data Landscape is Changing!"

- Automating Everything
- High Throughput
- Automation + Decisions = Autonomy
- Linked Data Is Foundational

Scientists are the worst judges of the reuse of their own data Review Article Published: 30 January 2023 The rise of self-driving labs in chemical and materials sciences

Accelerated 2DILM

Materials Discovery

NSF Data-Driven Materials Discovery Future Workshop, 2022

Achieve I-assisted and

Enable

data curation and community use

Predict

ynthesis onditions

Develop

Milad Abolhasani 🗠 & Eugenia Kumacheva

Nature Synthesis 2, 483–492 (2023) Cite this article

The high-throughput highway to computational materials design

Stefano Curtarolo 🖂, Gus L. W. Hart, Marco Buongiorno Nardelli, Natalio Mingo, Stefano Sanvito & Ohad Levy

Nature Materials 12, 191–201 (2013) Cite this article

43k Accesses | 1357 Citations | 34 Altmetric | Metrics

Review Article | Published: 09 March 2023

Combinatorial synthesis for AI-driven materials discovery

John M. Gregoire 🖂, Lan Zhou & Joel A. Haber

Nature Synthesis 2, 493–504 (2023) Cite this article

als Research

1258 Accesses | 5 Altmetric | Metrics

Data Alliance

Linking Data GEMD Graphical Model Yttrium Orthovanadate YVO₄



Gannon Murray PARADIM REU from Earlham College

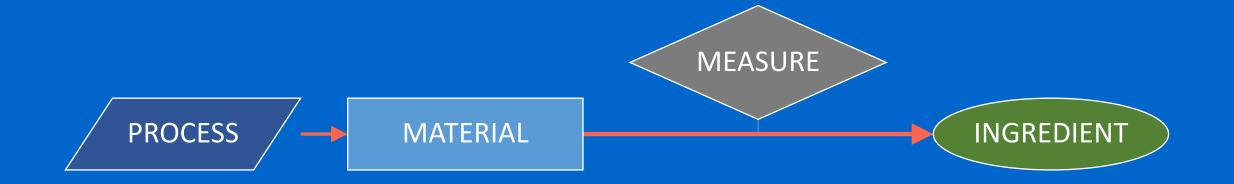




MaRDA Materials Research Data Alliance

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Graphical Expression of Materials Data (GEMD)



Data driven approaches are transformative but don't underestimate that: Connected data is unknown territory







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Creating a Material History



Gannon Murray PARADIM REU from Earlham College





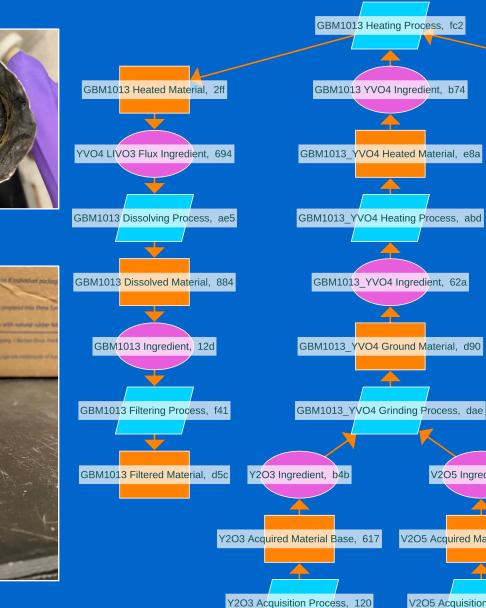


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GBM1013

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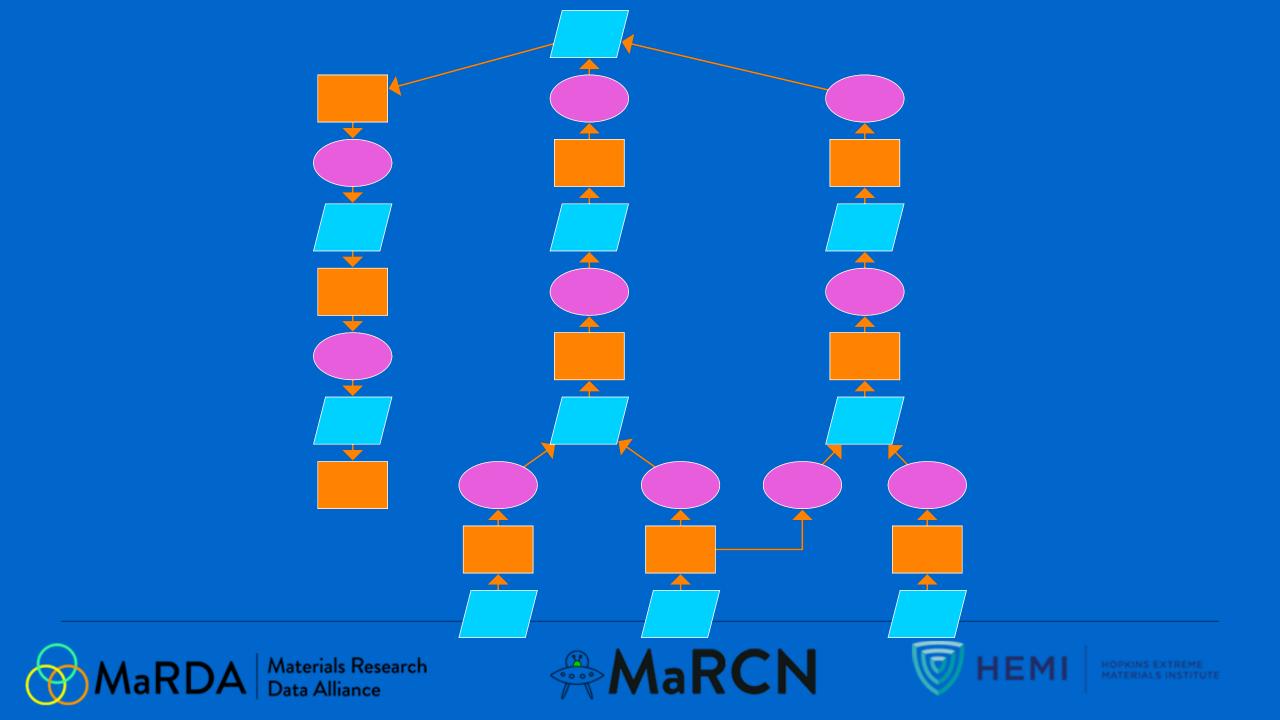
"Location, Hot Lab":"False" "Step 1, {'quantities': {'Duration': 0.0, 'Temp': 0.0, 'Type': 'Init'}, 'type': 'nominal composition'}":"False" "Step 2, {'quantities': {'Duration': 9.0, 'Temn': 900 no'}. 'type': GBM1013¹³Flux Medium Heated Material, 08a 'nomināl_c Faise "Step 3, {'quantities': {'Duration': 1.0, 'Temp': 900. ל'}, 'type': GBM1013 Flux Medium Heating Process, 5e1 uration': 9.0, "Step 4, {'q 'Temp': 0.0, 'Type'' 'Ramp'}, 'type': GBM1013_Flux_Medium Ingredient "Step 5, {'qu Jration': V2O5 Ingredient, 993 'Temp': 0.0, 'Type': 'End'}, 'type': V2O5 Acquired Material Base 'nominal_composition'}":"False" label:"GBM1013 Flux Medium H V2O5 Acquisition Process, Process, 5e1"



MaRDA Materials Research Data Alliance **MaRCN**



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MaRCN



HOPKINS EXTREME MATERIALS INSTITUTE

Event Driven Design

Seamless

Dropbox-like entry point

Automate Anything Repetitive

Curation Reduction **Active Learning**

Analysis Pipelines

Path to Deployment

Maintainable

Flexible (Future Foundation)



INTRO TO OPENMSISTREAM: Introduction to OpenMSIStream Installing OpenMSIStream Tutorials WORKING WITH OPENMSISTREAM: Using Main Programs Services/Daemons Encrypting Messages With KafkaCrypto Troubleshooting Support and Contribution **API REFERENCE/DEV INFO:** API reference Code Tests

OpenMSIStream

Search docs



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OpenMSIStream

OpenMSIStream

OpenMSIStream provides Python applications for generalized laboratory, analysis, and computational materials data streaming using Apache Kafka. More information is available on the Introduction page for new users. OpenMSIStream is maintained as part of the Open Materials Semantic Infrastructure (Open MSI), under NSF award #1921959, and has been published in the Journal of Open Source Software.

https://openmsistream.readthedocs.io/



OpenMSIStream: A Python package for facilitating integration of streaming data in diverse laboratory environments

Margaret Eminizer ^{1¶}, Sam Tabrisky^{2,3,4}, Amir Sharifzadeh ¹, Christopher DiMarco ⁶⁴, Jacob M. Diamond ⁶^{4,6}, K. T. Ramesh ⁶⁴, Todd C. Hufnagel ⁽⁶⁾ ^{4,5,6}, Tyrel M. McQueen ⁽⁶⁾ ^{4,5,7,8}, and David Elbert ⁽⁶⁾ ^{1,4}

1 Institute for Data Intensive Engineering and Science (IDIES), The Johns Hopkins University, USA 2 Department of Biology, Dartmouth College, USA 3 Department of Computer Science, Dartmouth College, USA 4 Hopkins Extreme Materials Institute (HEMI), The Johns Hopkins University, USA 5 Department of Materials Science and Engineering, The Johns Hopkins University, USA 6 Department of Mechanical Engineering, The Johns Hopkins University, USA 7 Department of Chemistry, The Johns Hopkins University, USA 8 Institute for Quantum Matter (IQM), William H. Miller III Department of Physics and Astronomy, The Johns Hopkins University, USA ¶ Corresponding author

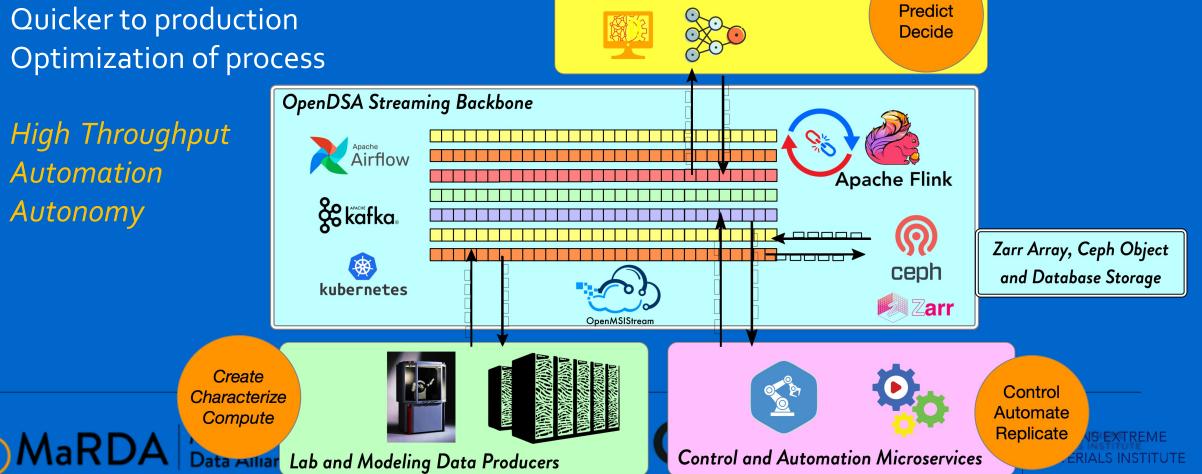
DOI: 10.21105/joss.04896

https://doi.org/10.21105/joss.04896

Event Driven Architecture makes Data the Unifying Thread that automates FAIR; accelerates analysis and AI/ML deployment; and empowers novel science and autonomy

Better Science and Production:

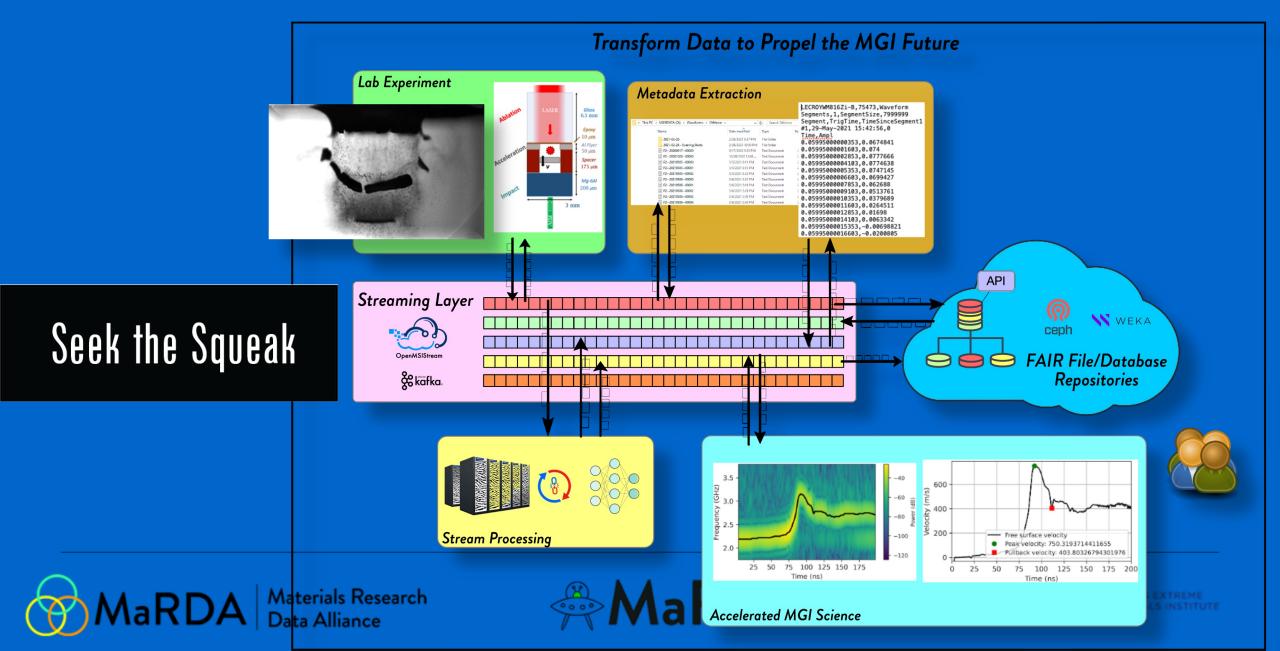
- Faster Discovery
- New Capabilities •
- **Quicker to production** •
- **Optimization of process** •



Data Science & Machine Learning

Learn

Example: DMREF Spall Resistant Aluminum OpenMSIStream

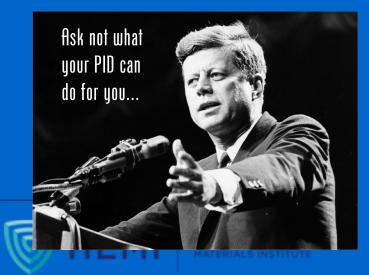


A Possible Path?

- PID Interoperability
 - empowers progress
 - perfect is the enemy of the good
- Outcomes Oriented PID as a Service
- Value Added Tools
- Seek the Squeak
- Lean Into Stakeholder Connections
 - Who controls the first mile?
 - Who controls the last mile?
 - Who gets the impact?
- MaRDA/MaRCN Connect Stakeholders

2023 2025 Do you know 2030 where your data are? 2035...

Got PID?



MaRDA Materials Research

