



Agenda

- Introduction
- Motivation
- Semantic Technology
- 5-Star Battery Data
- Examples!



Good Advice for Presentations

- Stay on time!
- Make the text as big as you think it needs to be... then make it bigger!
- Know your audience!



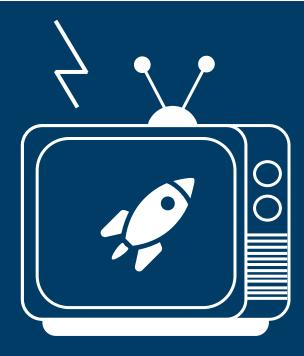
Data is available from the corresponding author upon reasonable request

Well-described, open datasets are some of the most valuable yet under-utilized resources we have in battery research.

Research articles that include data have higher impact via data re-use and citations.

I. Motivation

Computer, retrieve **engine data** and plot **new course**!



Science Fiction

Computer, retrieve the OCP curve for graphite!

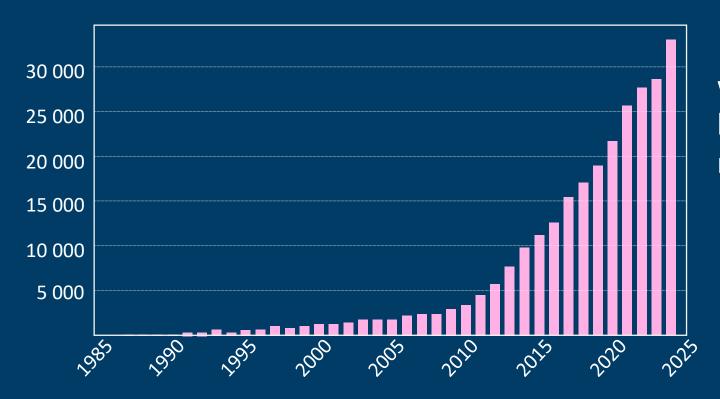


Science Fact (Almost)



The Knowledge Paradox

There have been more battery articles published in the last 5 years, than the preceding 30-year period

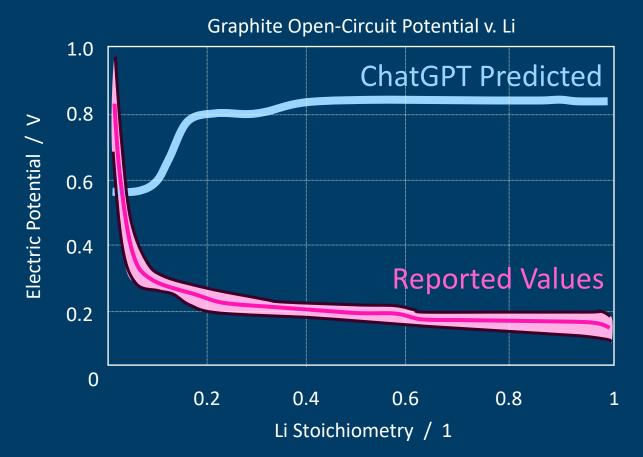


We have access to a wave of new information....but how much structured data is really behind it?



The Knowledge Paradox

Chatbots are excellent responding to queries about **stats data** from sports, finance, and pop culture ... but struggle with battery **data** queries. Why?



Structured, semantic data.

A cursory literature search* for Graphite OCP data that contains:

- ✓ Raw Data
- ✓ Open Format
- ✓ Permissive License
 Yielded 6 results. We need more.



The Knowledge Paradox

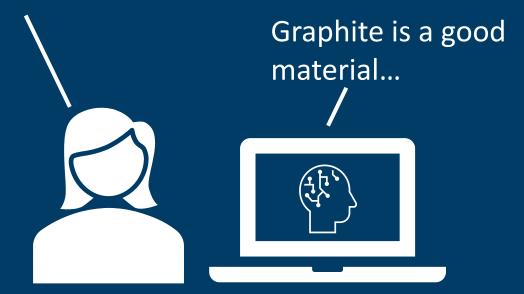
The bottleneck is **not data generation**... it's **knowledge extraction**.

We need to give data structure, meaning, and links.

II. Semantic Technology

Chatbots like ChatGPT are reactive and interact with the user through a flat interface

What is a good anode material for Li-ion batteries?



Agentic AI understands goals, makes plans, takes initiative, uses tools, and even interacts with other agents. It's designed to complete complex tasks.

Evaluate if our Si-Gr anodes are suitable to reach the cell KPIs for Gen 3b with LNMO.



This requires an awareness and understanding of the different systems that are available, the meaning of the data that they contain, and their exchange interfaces

2025



Agents in the Semantic Web

Scientific American: Feature Article: The Semantic Web: May 2001



The Semantic Web

A new form of Web content that is meaningful to computers will unleash a revolution of new possibilities

by TIM BERNERS-LEE, JAMES HENDLER and ORA LASSILA

SUBTOPICS:

Expressing Meaning

Knowledge Representation

Ontologies

Agents

The entertainment system was belting out the Beatles' "We Can Work It Out" when the phone rang. When Pete answered, his phone turned the sound down by sending a message to all the other *local* devices that had a *volume control*. His sister, Lucy, was on the line from the doctor's office: "Mom needs to see a specialist and then has to have a series of physical therapy sessions. Biweekly or something. I'm going to have my agent set up the appointments." Pete immediately agreed to share the chauffeuring.

"The Semantic Web is not a separate Web but an extension of the current one, in which information is given well-defined meaning, better enabling computers and people to work in cooperation."

Agent (n): A software entity that acts autonomously on behalf of a user or another system, using structured, semantically annotated data and web standards to make decisions and coordinate actions.



Agentic AI is powered by the intelligence of AI and the structure of the Semantic Web.



Agents in the Semantic Web

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"Artificial intelligence and Web researchers have co-opted the term for their own jargon, and for them an ontology is a document or file that formally defines the relations among terms. The most typical kind of ontology for the Web has a taxonomy and a set of inference rules."



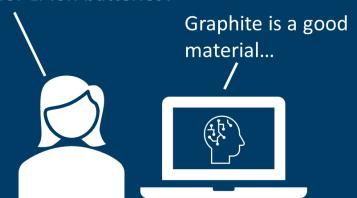
Ontologies have been extremely well developed for many general knowledge that enable search engine optimization. But they are slower to reach technical or scientific domains.



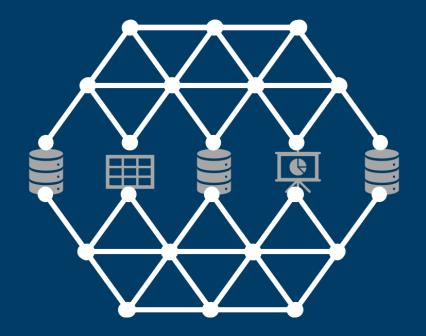
Agents in the "Battery Semantic Web"?

We have very good tools and infrastructure for humans to interact with machines.

What is a good anode material for Li-ion batteries?



But we are lacking the extensive networks of battery data and tools for agents to work with.





Semantic Technology

Methods and tools that attach meaning to data.

Linked Data

Data that is semantically annotated and linked to other pieces of data.

Knowledge Graph

An extension of an ontology that is used to describe actual instances of data and information. **Like a social network.**

Ontologies

A rich resource that includes relationships between terms to formalize knowledge about a domain. Like a textbook.

Controlled Vocabularies

Collection of terms with a defined and controlled meaning. **Like** a dictionary.

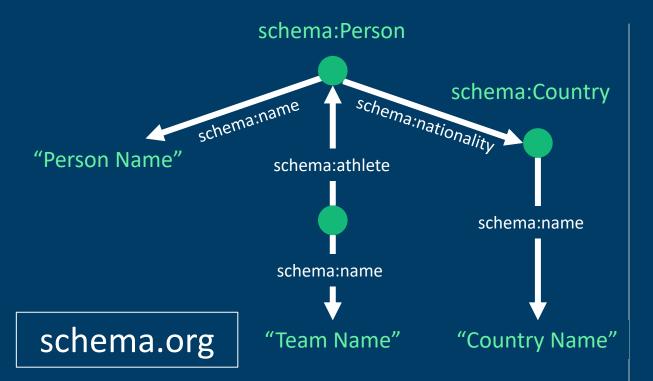
Word Wide Web Standards

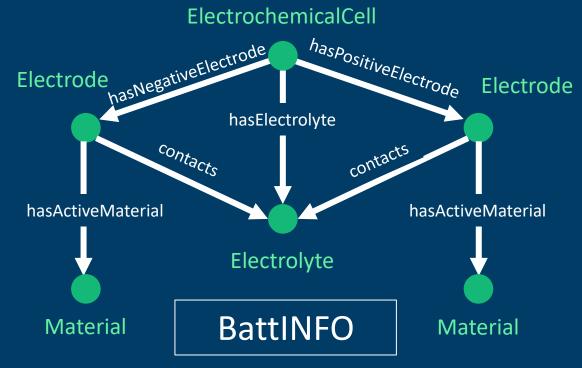
Sets the fundamental rules and grammar. Builds on "triples" like: Simon – isA- Person; Simon – works at – SINTEF



What is an Ontology?

A formal description of knowledge about a domain that is understandable for both people and machines.

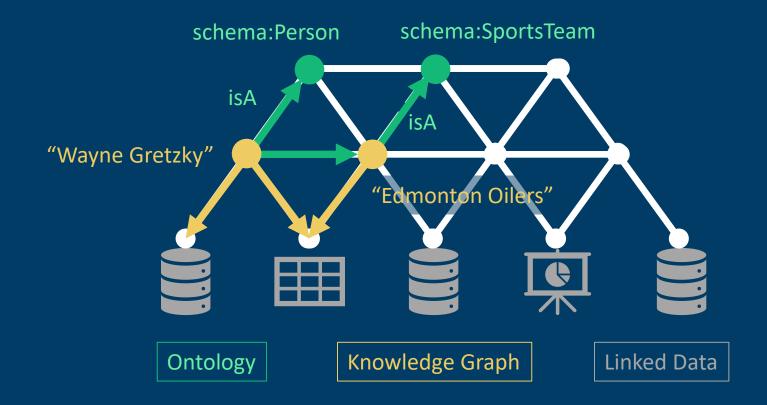






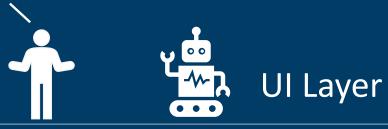
The Semantic Web

The Semantic Web is an extension of the World Wide Web, designed to be a web of data, navigated by machines.

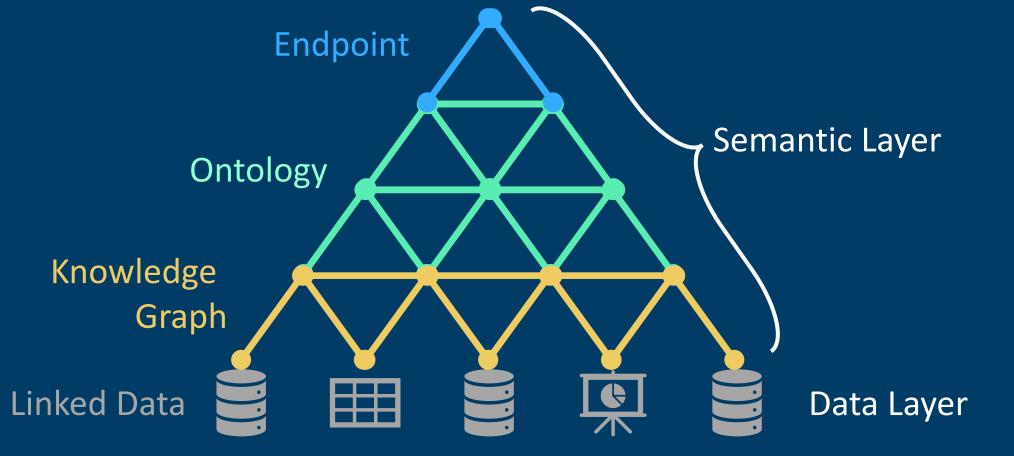


Computer, retrieve the results from discharging my CR2032 cell!













(do a search example)

III. Five-Star Battery Data





Five-Star Battery Data





***** Public Repositories and Permissive Licences

Goal: Ensures that battery data is stored in an accessible place and assigned a permissive license.

Public data repositories often offer additional benefits including:

- Managing persistent, unique, and resolvable identifiers (e.g. DOI)
- Supporting version control
- Structuring and embedding bibliographic metadata

Let's explore an example of a public data repository

(switch to Zenodo)





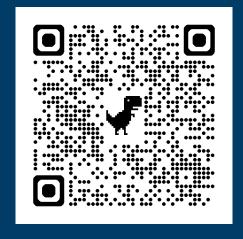
** Structured and Machine-Readable Data

Goal: Organize your data so that it has some clear regular pattern that defines the pieces and allows them to be read and processed by machines.

Common structures make it easier to process data.

- **Structured** means that your data is organized in a regular, defined, accessible structure (not just a blob of text)
- <u>Machine-readable</u> means that the structure and its content can be read by machines (like code, parsers, or other software tools) in a way that is indexable, unambiguous, and queryable

(switch to BDF)





** Structured and Machine-Readable Data

The SI and IUPAC style guidelines make some important statements about quantities:

- > Quantities are comprised of two parts: a value and a unit, separated by a space
- > Quantities follow the rules of algebra

When we instantiate a quantity like voltage:

Voltage = 4.2 V

We can isolate the value by dividing both sides by the unit:

Voltage / V = 4.2

That is why proper quantity notation in tables and plot axes include a slash before the unit, and not parenthesis or square brackets.



★★★ Open Data Formats

Goal: Provide your data in an open, non-proprietary format that anyone can access.

Battery data is often first serialized in proprietary binary file formats

- Makes it difficult to integrate that data with other systems or processing tools
- Acts as a barrier to sharing and re-use of research data
- Open data formats can make data more accessible and increase its value



★★★ Open Data Formats

A variety of open data formats for different uses. CSV is good when human-readability is important. Parquet is generally best for large, tabular datasets. JSON is suited for metadata. HDF5 supports more complex arrays.

Format	Structure	Human- Readable	Compressed	Typical File Size	Best For
CSV	Tabular	Yes	No	Large	Simple tables, small time series
PARQUET	Tabular	No	Yes	Small	Large time series, structured data
ТХТ	Unstructured	Yes	No	Large	Logs, raw data
JSON	Hierarchical	Yes	No	Medium	Metadata, structured data
HDF5	Hierarchical	No	Yes	Small	Arrays



★★★ Open Data Formats

If you use .CSV format and you are working in a country that uses the comma (,) as the decimal place indicator, please be sure to serialize with a dot decimal instead! It is very difficult to read files that use a comma for both the decimal and the delimiter!



Use a dot (.) as the decimal place

Test Time / s, Voltage / V, Current / A

1.00, 4.20, 0.50

2.00, 4.18, 0.50

3.00, 4.16, 0.50



Use a comma (,) as the decimal place

Test Time / s, Voltage / V, Current / A

1,00, 4,20, 0,50

2,00, 4,18, 0,50

3,00, 4,16, 0,50



★★★★Ontology-Annotated Metadata

<u>Goal</u>: Create machine-readable metadata using controlled, semantic vocabularies and ontologies.

Metadata provides essential information to interpret data and extract its full value.

- Type of battery test being carried out
- Type of battery or material being tested
- Quantities and units used in the data

Ontology-annotated metadata ensures that this information and its meaning is easily available to both humans and machines

DON'T PANIC

It's really the responsibility of software developers and data stewards to create tools that do this for you



★★★★ Ontology-Annotated Metadata

Unstructured text is the most common way of expressing metadata today.

This dataset contains test data obtained from cycling a lithium-polymer battery at 1C under ambient laboratory conditions. The result is tabular data, serialized in csv files with the following columns.

U(V): Cell Voltage

|I(A): Cell Current

•••



Difficult for machines to parse and understand. Structured metadata with controlled vocabularies is better.



★★★★Ontology-Annotated Metadata

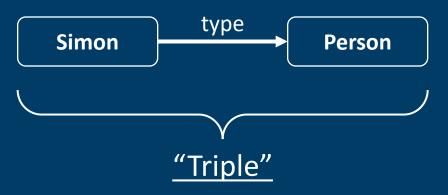
There are three important aspects to consider when creating ontology-annotated metadata:

- 1. <u>Structure</u>. Machines interpret metadata as a web-like graph of nodes that are linked by edges.
- **2.** <u>Terms</u>. Controlled vocabularies provide a dictionary of terms that can be used to create metadata.
- **3. Format**. Metadata is stored in file formats that can be read by both people and machines, like JSON-LD.



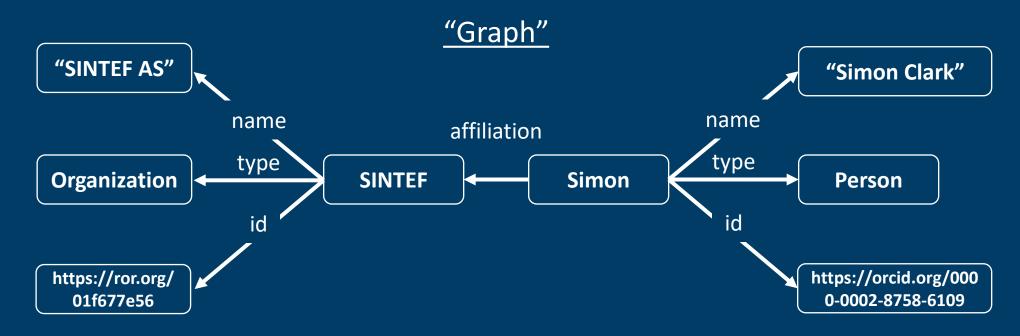
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2. **Terms**. Controlled vocabularies provide a dictionary of terms that can be used to create metadata.

RDF / RDFS

Foundational terms, that set the rules

DCAT

Terms for describing data catalogues

CSVW

Terms for describing tabular data on the web

Schema.org

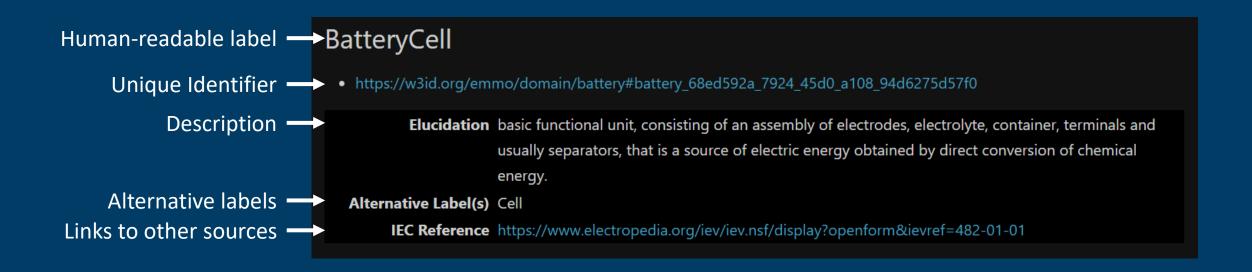
General purpose terms for search engine optimization

BattINFO

Terms for describing batteries, materials, tests, etc.



2. **Terms**. Controlled vocabularies provide a dictionary of terms that can be used to create metadata.





3. <u>Format</u>. Metadata is stored in file formats that can be read by both people and machines. JSON-LD is an easy serialization format for expressing RDF graphs in a human-readable way: metadata.jsonld

```
"@context": "https://w3id.org/emmo/domain/battery/context", ← Dictionary of terms
```



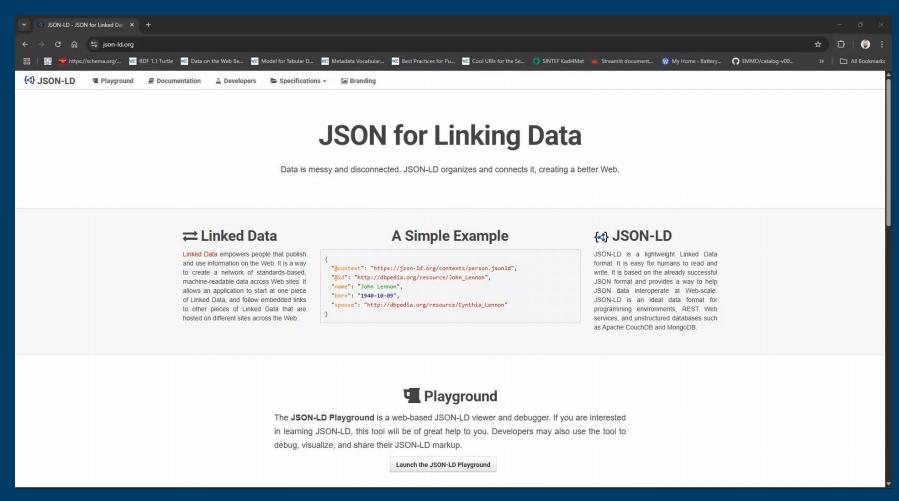
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Learn more at: https://json-ld.org/



```
"@context": "https://w3id.org/emmo/domain/battery/context",
    "@type": "BatteryTest",
    "@id": "urn:uri:bbdc6a42-550b-4038-a0a0-9236048b125f",
                                                                                            BatteryTest
    "hasTestObject": {
         "@type": "BatteryCell",
         "@id": "urn:uri:89afd7f5-b62f-4749-bba8-70fe4cffaad8"
                                                                              hasOutput
                                                                                                          hasTestObject
    "hasOutput": {
         "@type": ["dcat:Dataset", "BatteryTestResult"],
         "@id": "urn:uri:0d8197a5-6801-47e8-a45c-95f3d7b248dc",
                                                                                                          BatteryCell
                                                                                Dataset
         "dcat:distribution": {
              "@type": "dcat:Distribution",
              "@id": "urn:uri:2d610bd8-7a65-4ba8-8433-4c2ac23f96df",
              "dcat:mediaType": "text/csv",
              "dcat:downloadURL":
"https://zenodo.org/records/15069341/files/sintef__cr2032__discharging_11_mA__20240216.bdf.csv", "csvw:tableSchema": "https://w3id.org/battery-data-alliance/ontology/battery-data-
format/schema"
```









Three tips for implementation

1. Use BattINFO terms and structures in your metadata.

Many contain built-in mappings to other common data sources like Wikidata,
 Pubchem, etc.

2. Submit your dataset to the Battery Knowledge Base community on Zenodo

 We can scrape the metadata from datasets in the community to index them in the Battery Knowledge Base and link them with other datasets.

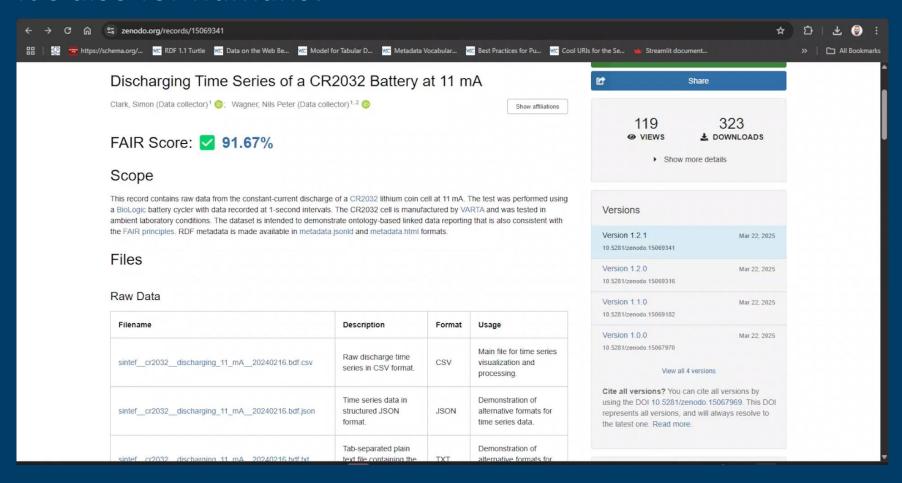
3. Make explicit links in your metadata to other datasets that are related

- schema:isBasedOn, schema:hasPart, schema:isPartOf
- Include human-readable links in the description



★★★★★Linked Data

It's also for humans!





Python Packages

Validation is like salt...not too little, not too much.

```
example custom cell.pv U X 👂 api reference.rst U
                                                  guick start.rst M
                                                                      installation.rst M
 examples > scripts > 💨 example_custom_cell.py > ...
  13 # Define the Bill of Materials
      lithium = helpers.merge models(cold.Lithium, cold.Foil)()
       mno2 = helpers.merge_models(cold.ManganeseDioxide, cold.Powder)()
       ec = cold.EthyleneCarbonate()
        emc = cold.EthylMethylCarbonate()
       lipf6 = cold.LithiumHexafluorophosphate()
       solvent = cold.Mixture(hasConstituent=[ec, emc])
       lp57 = cold.Solution(hasSolvent = solvent, hasSolute=lipf6)
       # Define material properties
       ec.hasProperty = [cold.MassFraction(0.3, "UnitOne")]
       emc.hasProperty = [cold.MassFraction(0.7, "UnitOne")]
       lipf6.hasProperty = [cold.MassFraction(0.127, "UnitOne")]
       # Build electrodes
       ne = cold.Electrode(hasActiveMaterial = lithium)
        pe = cold.Electrode(hasActiveMaterial = mno2)
  31 # Define casee
       case = cold.CoinCase()
       case.hasProperty = [
           cold.Diameter(20, "MilliMetre"),
            cold.Thickness(3.2, "MilliMetre")
  36
  38 # Build cell
             OUTPUT DEBUG CONSOLE PORTS JUPYTER

✓ TERMINAL

                                                                                                                              PS C:\Users\simonc\Documents\Github-local\DigiBatt\cold>
 PS C:\Users\simonc\Documents\Github-local\DigiBatt\cold>
```

- Flexible definitions to support describing complex objects
- Built-in validation for units
 - Hard validation that enforces correct unit types
- Built-in validation for ontology semantics
 - Soft validation that flags questionable usage
- Maintains a single-source of knowledge in the ontology







Excel-Based Interfaces: BattINFO Converter

B136 \checkmark \vdots \times \checkmark f_x \checkmark 2500 \checkmark					
1 2 3	A	В	С	D	i
	1 Metadata	Value	Unit	Type	
	2 Cell identification				
	3 Cell type	CoinCell	No Unit	string	
	Cell ID	Empa-bco-000007	No Unit	string	
	5 Date of cell assembly	16/6/2024	No Unit	string	
	Institution/company	Empa	No Unit	string	
	7 Scientist/technician/operator	Corsin Battaglia	No Unit	string	
	8 BattINFO CoinCellSchema version	Dev_Excel_for_v060	No Unit	string	Comment
	9 Project	Battery2030+/PREMISE	No Unit	string	Comment
	10 Assembled manually or by robot	manually	No Unit	string	Comment
	11 Positive electrode (cathode when battery is discharged)				
	12 Positive electrode current collector material	Aluminium	No Unit	string	hasPositiveElectr
	13 Positive electrode current collector thickness	15	um	float	hasPositiveElectr
	14 Positive electrode coating active material	Lithium Nickel Cobalt Manganese Oxide	No Unit	string	hasPositiveElectr
	15 Positive electrode coating active material chemical composition	LiNi0.6Co0.2Mn0.2O2	No Unit	string	hasPositiveElectr
	16 Positive electrode coating active material mass fraction	96	%	float	hasPositiveElectr
	17 Positive electrode coating binder material	PolyvinylideneFluoride	No Unit	string	hasPositiveElectr
	18 Positive electrode coating binder material mass fraction	2	%	float	hasPositiveElectr
	19 Positive electrode coating conductive additive material	CarbonBlack	No Unit	string	hasPositiveElectr
< >	Schema @context-TopLevel @context-Connector Ontology - Unit Unique ID +	1 •			•
Ready 92		Displa Displa	v Settinas 🏻 🖽		+ ± 160%



Community Building

Battery Semantic Technology is a community resource.

- Continue to develop the Battery
 Knowledge Base (BKB) as a humanreadable source of knowledge
- Monthly community "office hours" to answer questions and discuss improvements
- Widen the formal governance structure to cover a wider group of stakeholders

Teamwork makes the dream work!





Battery Semantic Technology

Never before have we had access to so much battery data, and the tools that we need to process it.

Semantic technology allows us to combine the power of machine processing with human intuition, for a new generation of battery innovation!



Summary

 Well-described, open data is a very valuable resource that can extend the impact of your work

 Use the Five-Star Battery Data guidelines when publishing your data

 Join our community discussions to help shape the future of battery data



Five-Star Battery Data





Thank You

SINTEF:

- Eibar Flores
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- Jesper Friis
- Francesca L. Blekken
- Casper W. Andersen

DTU:

- Tejs Vegge
- Ivano E. Castelli

EMPA:

- Corsin Battaglia
- Nukorn Plainpan

KIT:

- Christian Punckt



Fraunhofer ISC:

- Simon Stier
- Lukas Gold





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