Information extraction, dataset referencing and linked-data research at ACIS/U. Florida

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Outline

- Sample research projects @ ACIS
 - 1. Integrated Digitized Biological Collections (iDigBio)
 - 2. Self-aware Information Retrieval
 - 3. Data references and citations
 - 4. Linked data in the biological collections domain
- Conclusions

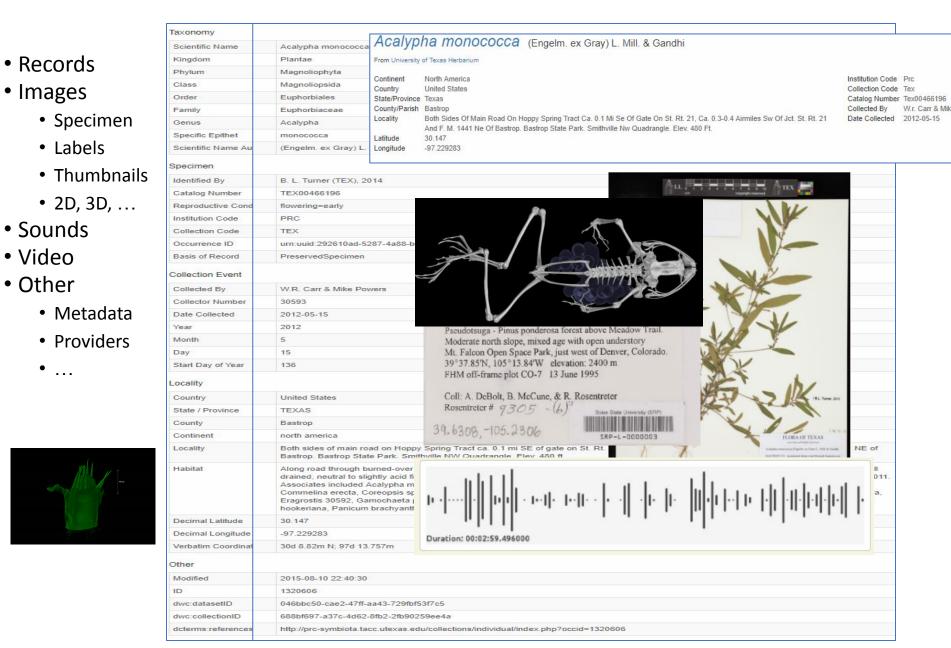
Digitization of Biocollections

- Information in biocollections can be used to understand environmental change, contaminants, biological invasions, disease transmission, and agriculture, among many other important areas.
- •There are about 1 Billion specimens in Biocollections in the USA and about 3 Billion in the whole World (Estimates).
- •NSF's Advancing Digitization of Biodiversity Collections (ADBC) program.

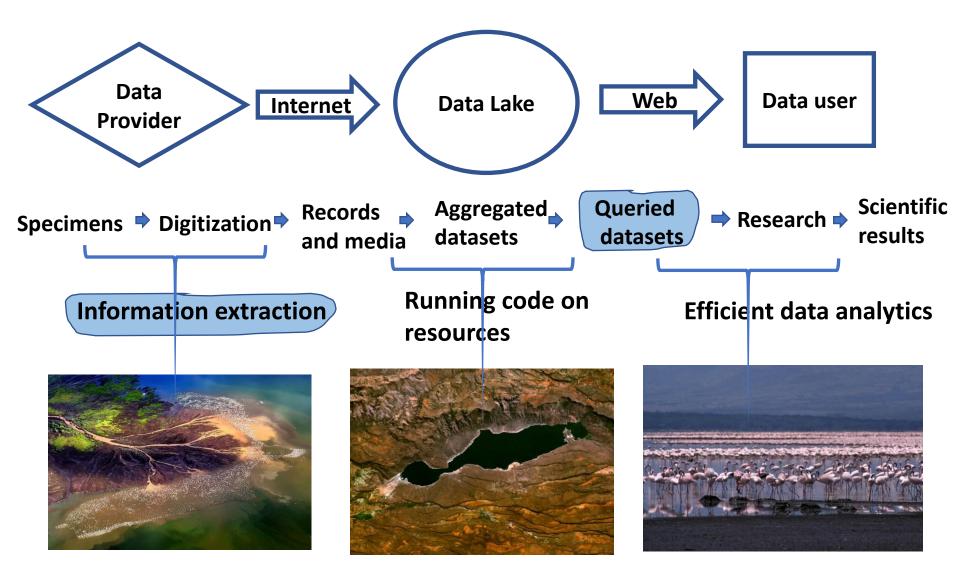


Photo by Chip Clark. U.S. National Herbarium at the Smithsonian Institution's National Museum of Natural History. Featured researchers: Dr. James Norris (right, front), research assistant Bob Sims (left, front), and associate researcher, Katie Norris (left, back).

Specimen Data: what are they?



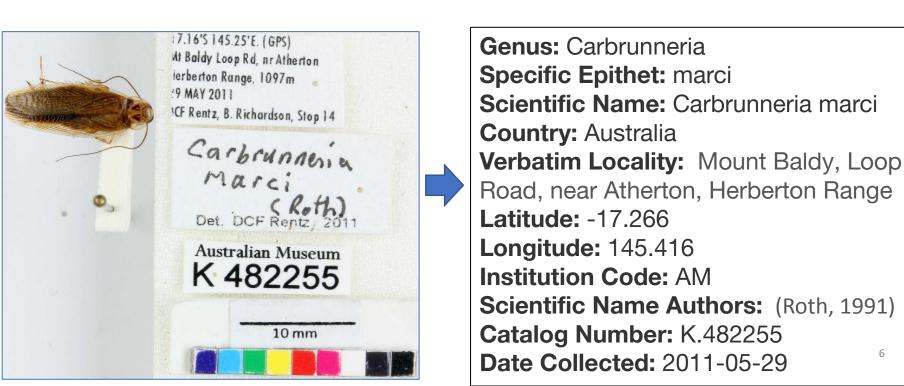
Basic Aggregator Cyberinfrastructure



Information Extraction (IE) Challenge

<u>Automated IE</u>: Optical Character Recognition (OCR) + Natural Language Processing (**NLP**)

- -Biocollections' images are problematic for OCR engines
- -OCR results are not perfect. Handwritten text is especially problematic.



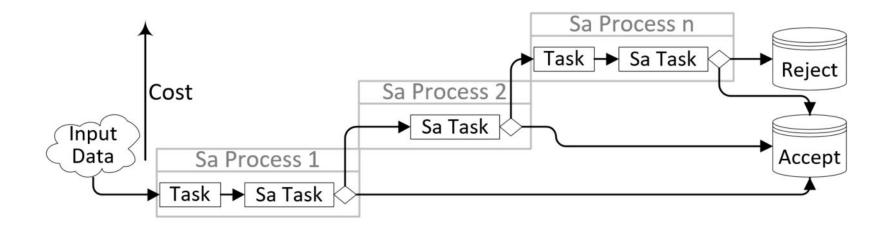
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Self-aware computational task (ChatGPT)

- A type of task performed by a computer system or artificial intelligence (AI) that involves the system's ability to monitor its own performance, analyze its own behavior, and adapt its own algorithms or strategies to achieve better performance or optimize its performance based on changing conditions or feedback.
- In other words, a self-aware computational task is one in which the system is not only able to perform a specific task or set of tasks, but also able to reflect on its own performance, identify potential areas for improvement, and modify its own behavior or processes accordingly. This level of self-awareness is typically achieved through the use of advanced machine learning and artificial intelligence algorithms, which enable the system to learn from experience and adjust its own behavior based on that learning.

SELFIE: Self-aware Information Extraction

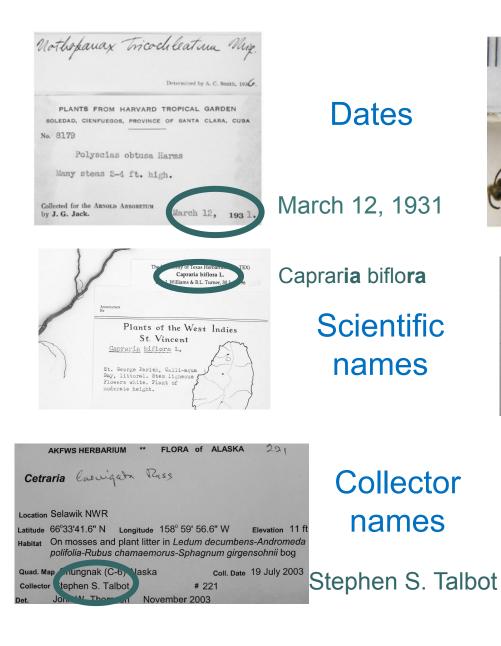
 Workflow of Self-aware Processes (SaP) consisting of Self-aware Tasks (SaT) and possibly other tasks



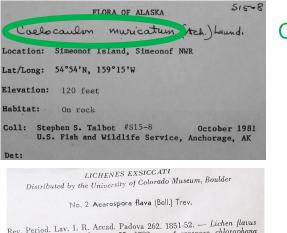
Self-aware Task (SaT)

Part	Input	Adaptable Script/program	Adaptable Acceptance Method	Outputs
Example	lmage x	/path/script1.py	[0,b) -> Task y [b,1] -> Accept	lmage x Value, Confidence

Information Extraction Examples







Rev. Period. Lav. I. R. Accad. Padova 262. 1851-52. — Lichen flavus
Bell., Append. Fl. Pedemont. 55. 1792. — Acarospora chlorophana
(Wg.) Mass., Ricerche 27. 1852. — A. oxytona (Ach.) Mass., ibid. p. 28. — A. sulphurata Arn., Lich. Ausl. 22: 63. 1886. — A. hilaris (Duf.)
Hue, Morph, et Anat. 113. 1909. — A. texana Magn. in Mycologia 21: 250. 1929. — A. gobiensis Magn., Monogr. Acar. 98. 1929. — A. novo-mexicana Magn. in Medd. Goteb. Tradg. 5: 58. 1930. — A. incertula mexicana Magil, In Medil, Goleb, 11adg. 5: 50, 1950, — A. incernua Magn., 1.c. p. 61, — A erythrophora Magn., l.e. p. 60, — A. weldensis Magn. in Goteb. Kgl. Vet. Handl. F. 6, ser. B, Bd. 6, No. 17: 15, 1956.

U.S.A. COLORADO. BOULDER COUNTYront Range 2 miles N.V base of Steamboat Mountain W. A. Weber, R. A. Anderson & D. D. Awasth 1,900 m. 1 Feb. 1961

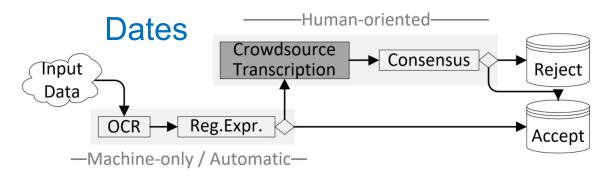


talus blocks, west

Caelocaulen muricat**um**

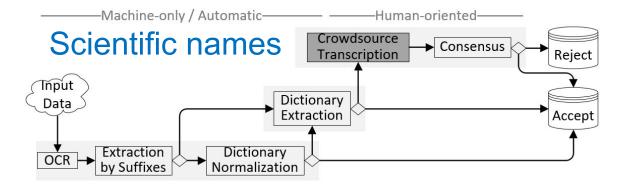
W. A. Weber, R.A. Anderson & D. D. Awasthi

SELFIE Workflow Examples

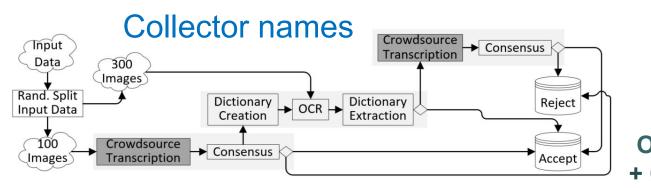


OCR: OCR software generates text with all information in the image.

Reg. Expr.: analyzer returns earliest date amor the "long" dates.



Extraction by Suffixes + Dictionary Normalization Or/followed by All text scan + Dictionary Extraction



OCR + Consensus + Dictionary Creation followed by OCR+ Dictionary Extraction + Crowdsource + Consensus

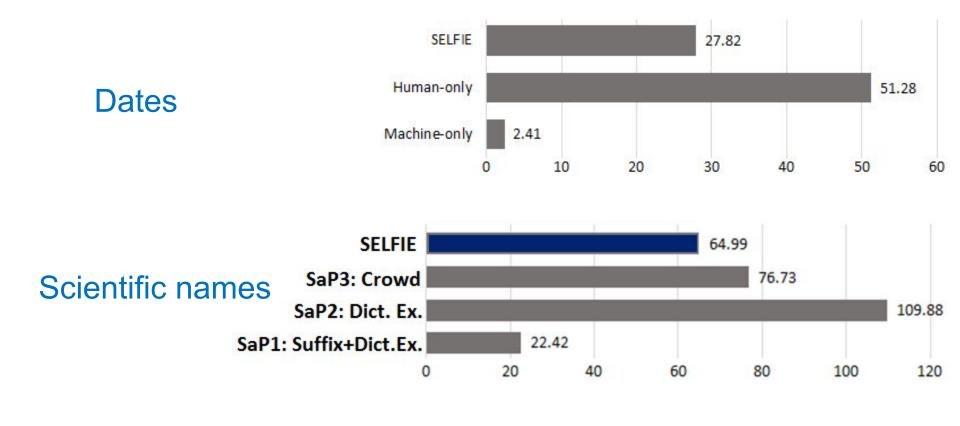
Similarity to expert transcription

	SaP/SELFIE	# Accepted	Similarity	SEM	Std. Dev.
Dates	Machine-only	48	0.934	0.024	0.167
Dales	Human-only	51	0.971	0.022	0.155
	SELFIE	99	0.953	0.016	0.162
	SaP/SELFIE	# Accepted	Similarity	SEM	Std. Dev.
Scientific	1. Suffix+Dict.Ex.	15	1.0	0.00	0.00
	2. Dict. Ex.	10	1.0	0.00	0.00
names	3. Crowd	66	0.944	0.026	0.214
	SELFIE	91	0.959	0.019	0.183

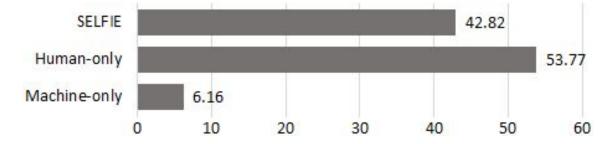
Collector names

1. Human 100i	92/100	0.900	0.030	0.000
		0.000	0.030	0.288
2. Machine-only	94/300	0.862	0.027	0.262
3. Human 300i	191/206	0.900		
SELFIE	375/400	0.895		

Seconds per accepted item



Collector names



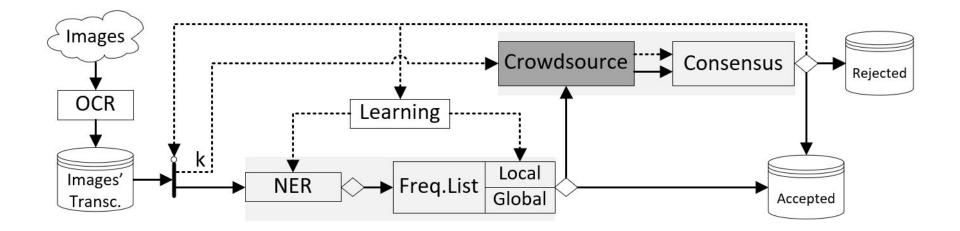
General Extraction of Candidate Values

Optical Character Recognition

- 1. The entire text of the images is extracted (Google Cloud OCR engine was used).
- Named Entity Recognition (NER)
 - •Challenge: **Train** the recognition model (**training values**)
 - Algorithm for Automated Labeling of DC Terms from Crowdsourced Data:
 - 2. Crowdsourced values are searched in the sentences of the OCR-ed text of their correspondent images.
 - 3. Training sentences are prepared (start end positions of every term in the sentence).
 - •The NER model is trained with the training sentences.
 - 4. Use the training set of sentences to customize NER model in the NLP library (spaCy was used). 13

Human-in-the-Loop Workflow for DC Terms Extraction

- Iterative training of the NER model.
- When no previous data is available or useful for a specific term.

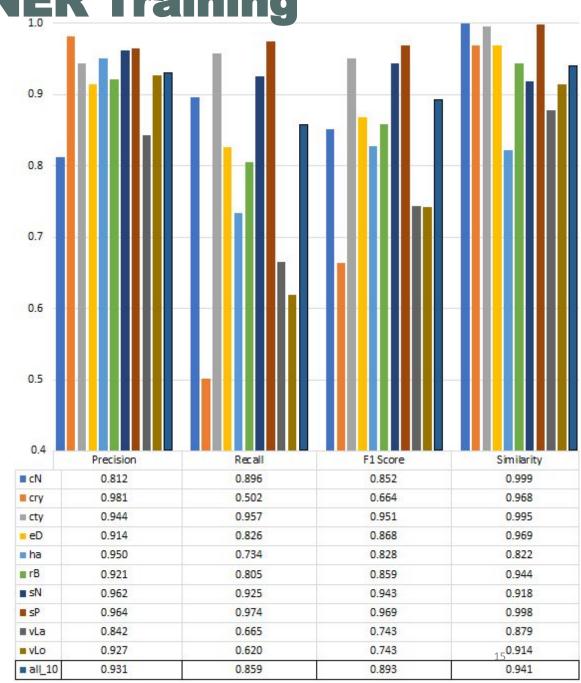


Experiments – NER Training

 The trained NER model was able to extract ~ 86% of <u>ALL</u> 10 DC values with a similarity to the ground-truth data of 0.931

Darwin Core terms:

- cN: catalogNumber
- cry: country
- cty: county
- eD: eventDate
- ha: habitat
- rB: recordedBy
- sN: scientificName
- sP: stateProvince
- vLa: verbatimLatitude
- vLo: verbatimLongitude
- **all_10**: weighted average for all the terms



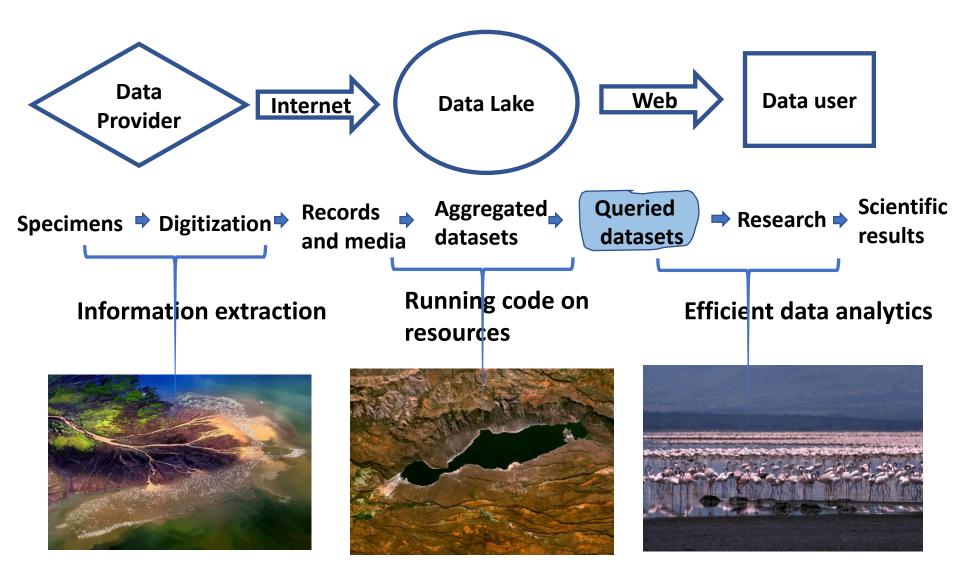
Confidence for Candidate Values

- •Global Frequency List (~static)
 - •Per-term number of times every value has repeated.
 - •Created with the data of iDigBio: more than 120 Million records.
- Local Frequency List (~dynamic)

 Created from specimens already processed in the biocollection under study

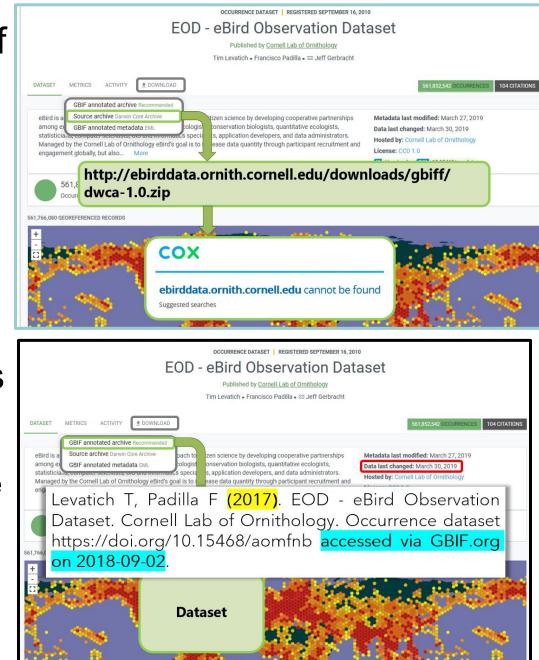
- •Algorithm (candidate_value, local_list, global_list):
 - •If repetitions(candidate_value, local_list) >= 3:
 - return(Accept)
 - If repetitions(candidate_value, global_list) >= 20:
 return(Accept)
 - return(Reject)

Biodiversity CY 1.0: Biocollections data lake



Reliable references and citations

- A reference is **reliable** if both
- Allows continued access to what is referenced
 - No link rot
- Only identifies what is referenced
 - No content drift/re-use
- Problem: How to reliably reference datasets served at



Content-based identifiers

 Cryptographic hash functions can produce unique content-based identifiers for digital datasets

Content-based identifier for the 2017 eBird dataset:

hash://sha256/<mark>29d30b566f924355a383b13cd48c3aa2</mark> <mark>39d4?cba0/ 55f4ccfc2930289b88b43c</mark>

Content-based identifier for the 2019 eBird dataset: hash://sha256/<mark>ec3ff57cb48d5c41b77b5d1075738b40</mark> f598a900e8be56e7645e5a24013dffc4

Signed citation

 a customary citation extended to include a content signature of the cited digital content



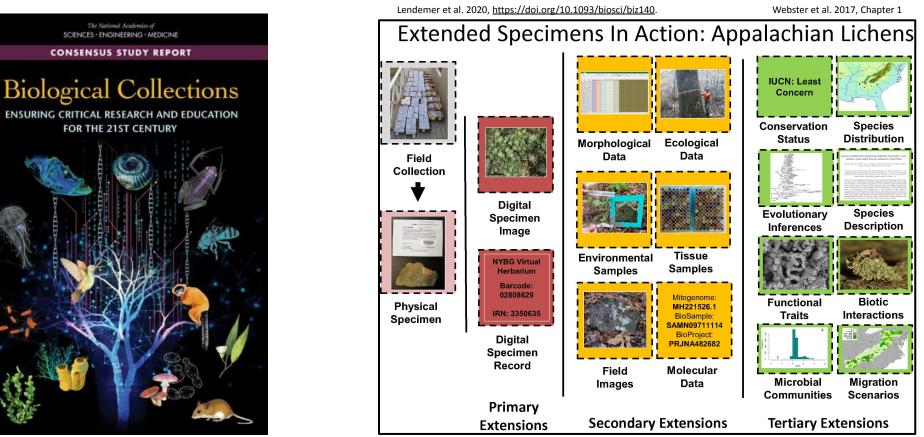
Museum of Comparative Zoology, Harvard University. 2021. Head Frontal View of MCZ:ENT:17219 Nomadopsis puellae (Cockerell, 1933) Accessed at http://mczbase.mcz.harvard.edu/specimen_images/entomology/large/MCZ-ENT_00017219 Spinoliella puellae hef.jpg on 2021-12-07.

3230808ea356

Benefits of signed citations

- 1. Verification: signatures can be regenerated
- **2.** Unique identification: hashes are statistically guaranteed to be unique
- **3.** Content-based: vs. location/publication based
- **4.** Decentralized resolution: registries & repositories can be created by anyone anywhere
- **5.** Robust resolution: rot detected, recoverable from (other) existing registries and/or copies
- **6.** Recursive citations: citations of collections of citations ... and more

Extended Specimen Concept



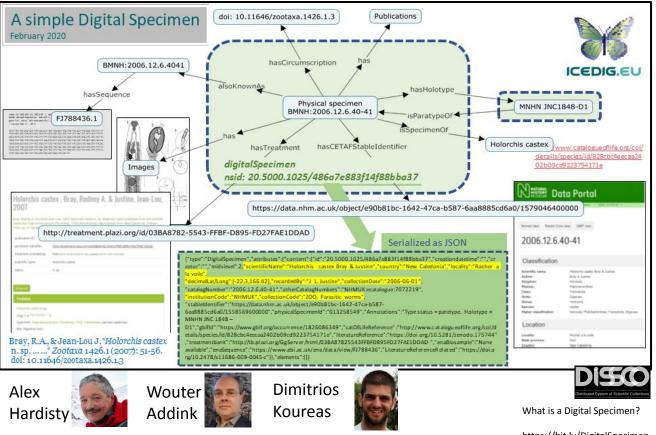
recommends building a network of extended specimens to facilitate research across taxonomic, temporal, and geospatial scales. ...the "holistic" (Cook et al., 2016) or "extended specimen" concept (Webster, 2017)

Extended specimen "a constellation of specimen and data types that, in combination, capture the multidimensional phenotype (and genotype) of an individual."

The National Academies of SCIENCES · ENGINEERING · MEDICINE CONSENSUS STUDY REPORT

FOR THE 21ST CENTURY

Opportunity: Richer IT Abstractions



Digital specimen "a

surrogate in cyberspace for a specific physical specimen, identifying its actual location and authoritatively saying something about its collection event (who, when, where) and taxonomy (what), as well as providing links to high-resolution images. A digital specimen exposes supplementary information about related literature, traits, tissue samples and DNA sequences, chemical analyses, environmental information, and much more, stored elsewhere than in the natural science collection itself."

https://bit.lv/DigitalSpecimen

FAIR Digital Objects (FDOs)

- •<u>F</u>indable <u>A</u>ccessible <u>Interoperable and <u>R</u>eusable</u>
- •FDO: a FAIR self-contained, typed, machine-actionable data package
 - Has a unique identifier
 - •Can be discovered, accessed, moved, replicated and managed individually and programmatically

Genomic data Biochemical data Morphological data Geographical data Ecological data Taxonomic data Related species

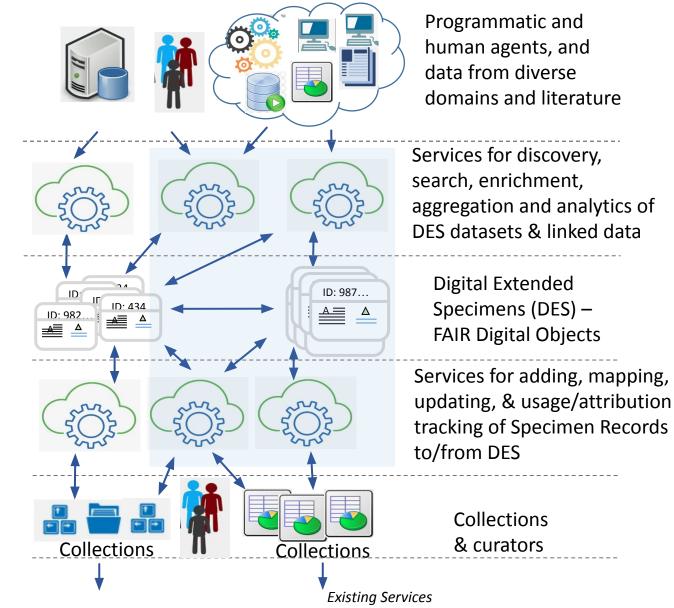
> Metadata Data operations

> > APIs

FDO Cyberinfrastructure Architecture (high-level)

FAIR objects and services

- Hosted locally/regionally by biocollections IT organizations (e.g. the unshaded mid-three layers) OR
- Hosted by global/community-level serving entities like today's "aggregators" (e.g the shaded area).



Acknowledgments

- National Science Foundation's Advancing Digitization of Biodiversity Collections Program [DBI-1115210 (2011-2018) and DBI-1547229 (2016-2021)
- iDigBio/ADBC Community and Collaborators
- Current ACIS IT team















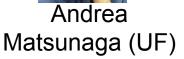
Renato Chris Wilson Maureen Randy Jesse Dan Stoner Michael Figueiredo (Software Kelly Fischer Bennett (Software Elliott (Professor) Developer) (Developer) (Developer) (Developer) Contractor) (PhD Student)

SELFIE work done with



Icaro Alzuru (UF)







Mauricio Tsugawa (UF)