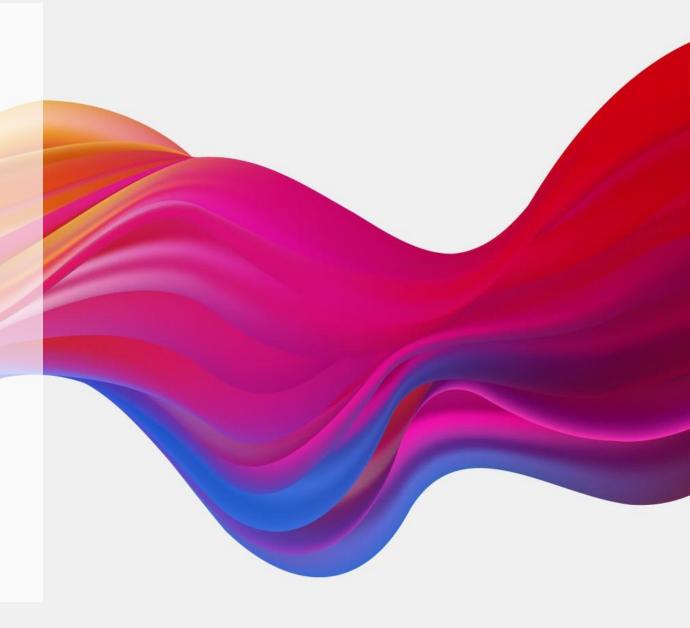
Building an **Analysis-Ready** Data Resource Using a Multi-Stage Pipeline

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MSDE/MSDS Practicum 2
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August 21, 2025







Introduction

Background, Objectives, and Importance



Project overview

Retrieve, Cleanup, and Enhance



Conclusion

Limitations and Future Improvements









What: Collaborative network of researchers

Why:

- Forum for data practitioners to come together, share their knowledge, and learn from one another
- Advance understanding of earth systems through use of data, information, tools, and techniques

How:

- Collaboration Areas: Groups formed around common interests that help address challenges and identify solutions that enable data integration efforts.
- Annual CDI Request for Proposals: Seed funds awarded each year for projects that focus on data integration.
- **Biannual Workshops:** In-person workshop where participants present their work, propose collaborative paths to solve data challenges, and share knowledge with network of peers.

An Emoji Evolution of Project Objectives









Why is Data Integration Important?

"Increasingly, data scientists must first sort through heterogeneous, incongruent, and fragmented datasets before any analyses can be conducted. Such problems with data availability are often exacerbated in emergency situations where real time analyses are often stymied by unevenly documented or unclean data."

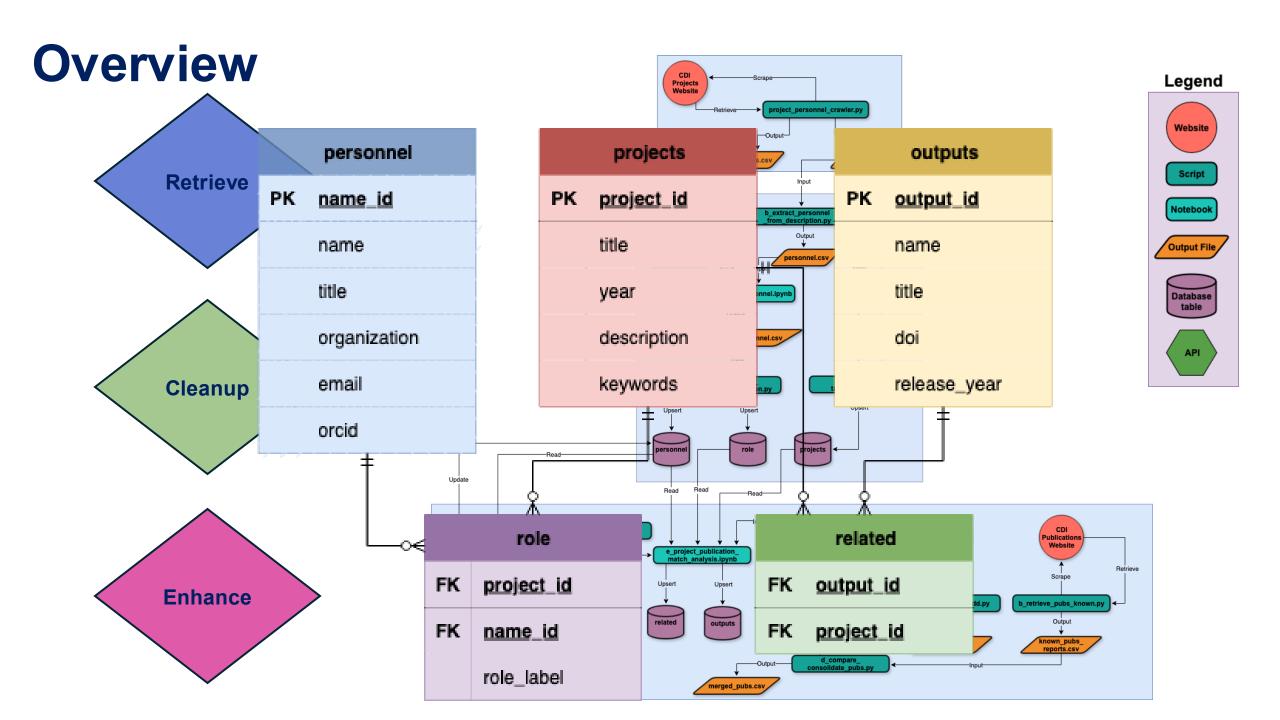
(Cheng, C. et al, 2024)

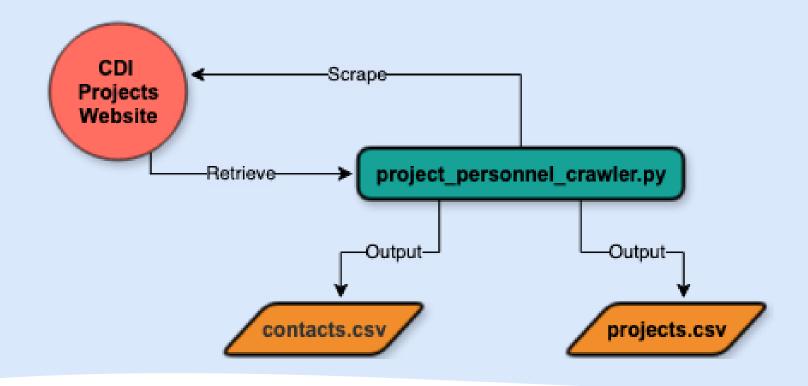
"One of the great challenges of data-driven research is that data rarely come in a form that is immediately ready for analysis." (Huber, R. et al, 2021).

"Integrating structured, semistructured, and unstructured data is crucial for insights and decisions. However, complexities like format mismatches, data quality issues, and interoperability hinder efficient data utilization" (Mirza, F., 2021)

"Data integration solutions leveraging data-driven techniques improve datadelivery capabilities and facilitate data analytics." (Liu, D. and Yoon, V., 2024)

"Integration of existing information systems is becoming more and more indispensable in order to dynamically meet business and customer needs while leveraging long-term investments in existing IT infrastructure." (Ziegler, P., and Dittrich, K., 2007)





Retrieve

- Scrape full-text content from CDI project web pages
- Extract title, year, description, and contacts
- Export structured project and project contacts (personnel) data

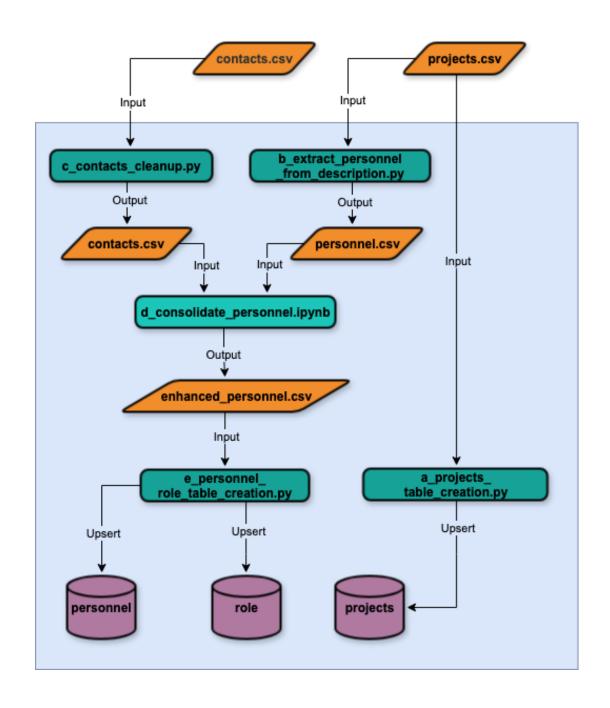
Cleanup

Projects: clean descriptions, assign unique identifiers, extract top-5 keywords, upsert into projects table

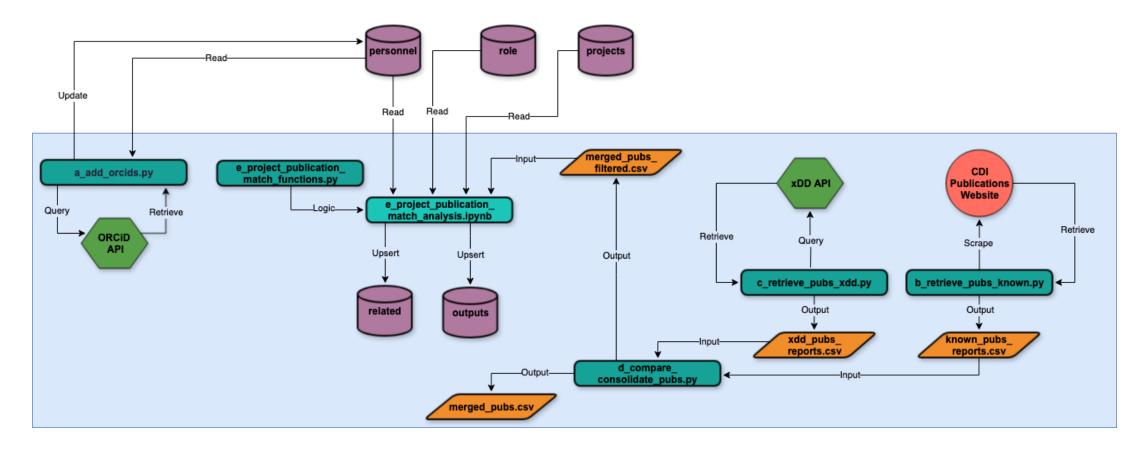
Personnel: regex-parse and clean names, tag lead PI versus collaborators, and standardize entries

Consolidation: merge personnel information from contact sections and descriptions, fuzzy-match name entries, select representative personnel entries, and remove duplicates

Final Load: assign each personnel entry a unique identifier, upsert into personnel and role tables in database



Enhance



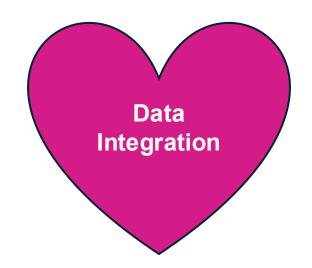
- ORCID enrichment via ORCID API
- Known publications retrieval from CDI
- xDD API publication extraction

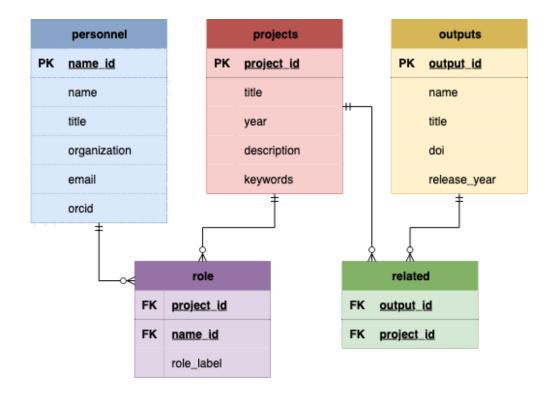
- Publications consolidation and filtering
- Project-publication matching and upsert

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Limitations

- Missing and incomplete publicly accessible CDI project data
- Project personnel may not have ORCIDs or have different names associated with their ORCIDs
- xDD is not a comprehensive corpus of all published works by CDI project personnel
- Project personnel may not be acknowledging the CDI or have cited the CDI inconsistent ways in their outputs
- Heuristic matching errors
- Fixed database schema may require refactoring if new data types are included in the future

















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