Data Management SOP for the Tampa Bay Estuary Program

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Prerequisites

Background

2.1 Importance of data

- Data are the foundation of all research products and management decisions
- A data definition, e.g., raw information in flat files, synthesized/derived datasets, models, etc.
- How data are used in applied research/environmental sciences

2.2 Why we need to effectively manage data

- What happens when data are not managed properly Figure 1, bit-rot
- Professorware
- Benefits of a data management workflow
- Applications in open science

2.3 Goals/objectives of this document

- What it is, what it is not including what makes TBEP different from other organizations, i.e., we have hands in lots of projects vs one central product (e.g., OHI), so our SOP needs to be generalizable
- Develop pathway for metadata
- Intended audience

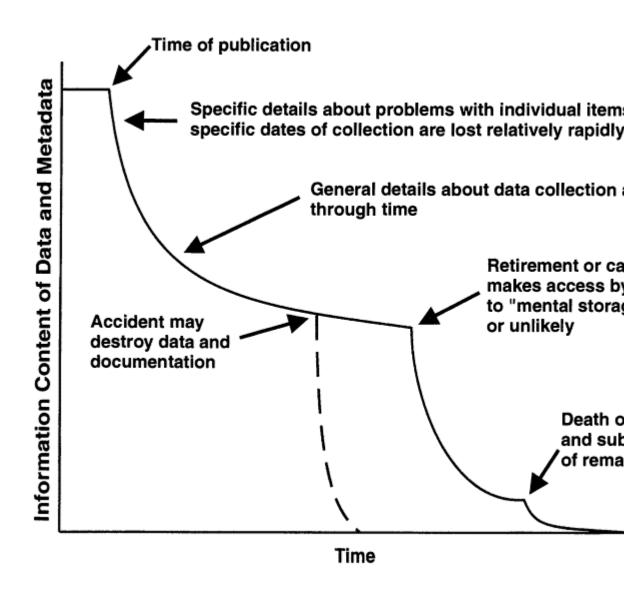


Figure 2.1: Loss of information over time in the absence of data management $[@{\rm Michener}97]$

Key Concepts and Principles

3.1 General concepts

- What are data, i.e., from the perspective of the researcher/agency scientist/manager?
 - Workflow (e..g, operationalizing Twitter scraping), dataset (field/lab data), model products, etc.
 - Ask yourself, who is going to use this and how do I make their lives "easier" by opening the data using FAIR principles?
- The FAIR principles (very broad, emphasize throughout), also general open science definition and how data relates to open science (channel PeerJ paper distinction)

3.2 Specific concepts

- Specific concepts (particularly for tabular data)
 - an overview of tidy data, can a machine read it?
 - normalized tables (including discussion of key variables), what are unique ids (e.g., therf oyster, how did I make the unique id?), facilitate standard DB queries
 - metadata documentation (min requirements, relevant standards)
- types of data products (e.g., raw data, models, synthesized/derived data, etc.) or types of data (flat file, spatial, disparate)
- how-to cookbooks for data prep (could speak to different parts of the analysis workflow, e.g., project inception, mid-project, post-project/damage control) for archival, naming conventions (e.g., no spaces, short but descriptive, etc.), data dictionaries

- where do data live long-term, what's a doi, considering a data paper, federated repository, etc.
 - GitHub repositoryStable URL

 - Official repository

Data Management Workflow

- Setup some kind of flow chart (if this, then that)
 - What type of project am I working on?
 - What types of products am I expecting?
 - Where am I at with the project (beginning, middle, end/damage control)?
 - How do I want to make the data accessible?

Case Studies

Demonstrate the workflow

- 5.1 TBERF oyster restoration project
- 5.2 DeSoto/RESTORE project
- 5.3 Red Tide Twitter repo

Final Words

- emphasis on "something is better than nothing", fully open is ideal but difficult to achieve
- Just remember FAIR

Appendice

7.1 Guidelines for best data management practices

7.2 Data types

- Field data
 - Survey forms
 - Tabular*
 - Database of tables
- Database
 - Synthesis
 - New
- Model
 - Actual model
 - Model results
- Dashboard

7.3 Definitions

Dashboard Data Aggregation vs. synthesis Model Tabular Database Flat file Tidy data

7.4 Metadata templates

General - Who, what, when, where, why?

Specific - XML, EML, etc.