

Best Practices for Storing Your Data

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Overview

Motivation

- Reduce the risk of data loss
- Speed up your research
- Increase visibility and collaboration
- Be kind to your future self

Overview

Contents

- General advice
- Storage strategies
- Hands-on
- Slides and code will be available

General Advice

General Advice

Storage data at the right place

- Raw data
- Intermediate data
- Data product
- Archival data

General Advice

Storage data at the right place

- /home: source code, executables, etc.
- /scratch: job input / output
- /projects: final job output (long term)
- /tigerdata: long term storage and for sharing

General Advice

Storage data at the right place

- TigerData
 - Requires sponsorship
 - Not available on the compute nodes
 - Organized by projects
 - <https://tigerdata.princeton.edu/how-it-works>

General Advice

Princeton Research Data Service (PRDS)

Data Services

- Review **data management plans** (DMPs)
- Advise on federal **funder requirements**
- Consult on file naming conventions and data organization
- **Data curation** assistance (no matter which repository you use!)
- Assistance preparing data for **publication**
- Help find an appropriate **repository**
- Help with **data documentation** and **READMEs**
- Mint **DOIs** for datasets
- **Large data** support (Globus access point)
- Hosts [**Princeton Data Commons**](#)



**Princeton Research
Data Service**

Scan to make an appointment

prds@princeton.edu

<https://researchdata.princeton.edu/>



General Advice

Data safety

- Storage data at 3 places (one original and two copies)
- Two types of storage media (e.g. local disk and Princeton cluster)
- One copy off-site (e.g. cloud storage or external data centers)

General Advice

Scalability

- Be prepared for larger dataset in the future
- Too many files can crash the file system
- Keep track of storage usage
- Compress data before archiving
- Discard easily reproducible data

General Advice

Reproducibility

- Keep raw data safe
- Document processing steps
- Save processing parameters alongside data
 - Software version may be important
- Append data instead of overwriting if possible
- Perform tests

General Advice

Shareability

- Use mainstream format if possible
- Bundle small files
- Split large files
- Document non-standard notations

Storage strategies

Storage strategies

Formats to consider

- Text
 - CSV - Flat tabular data
 - JSON - Nested metadata, parameters
- Binary
 - Raw binary - Temporary inter-process data
 - NPY / NPZ - Array data
 - HDF5 - Large hierarchical data
 - ADIOS2 - Exascale data emphasizing parallel writing

Storage strategies

Formats to consider

- There is not a format that is best for all use cases
- Understand your needs and choose the right format for your project

Storage strategies

Formats to consider

- There is not a format that is best for all use cases
- Understand your needs and choose the right format for your project
- Case study: SPECFEM++

Storage strategies

Choosing a default format for SPECFEM++

- SPECFEM++ is a C++ numerical simulation software for elastic wave
- We need to choose a default format for storing simulated wavefield
- Requirements:
 - Standard
 - User-friendly
 - Scalable

Storage strategies

Choosing a default format for SPECFEM++

- Text?
 - Not a consideration for large data

Storage strategies

Choosing a default format for SPECFEM++

- HDF5?
 - Widely supported
 - Good for large data

Storage strategies

Choosing a default format for SPECFEM++

- HDF5?
 - Widely supported
 - Good for large data
 - Parallel write issues with a large number of processors
 - Metadata syncing is slow
 - Not always easy for users

Storage strategies

Choosing a default format for SPECFEM++

- ADIOS?
 - Smoother installation process
 - Good parallel write performance

Storage strategies

Choosing a default format for SPECFEM++

- ADIOS2?
 - Smoother installation process
 - Good parallel write performance
 - Official documentation is not so good
 - Not a necessity for most users
 - Limited ecosystem (C / MatLab / Python / Fortran)

Storage strategies

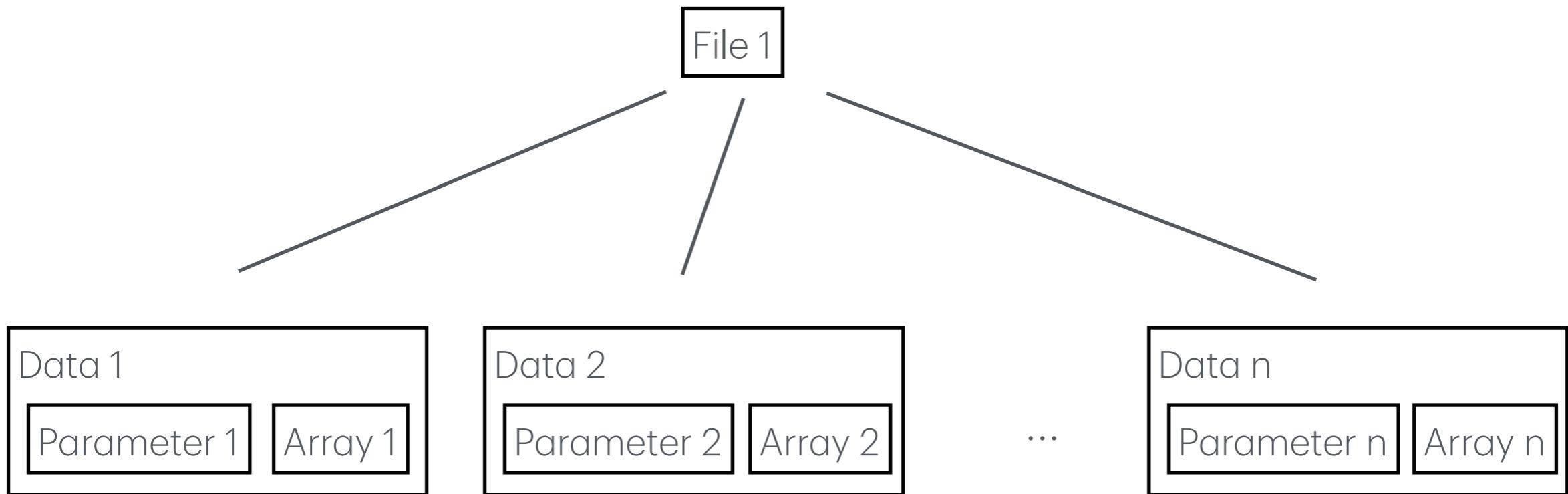
Choosing a default format for SPECFEM++

- NPY / NPZ
 - Simple - just raw binary + a small header
 - Widely supported (Not just Python)
 - Few dependency required (only zlib for NPZ)

Storage strategies

Separate parameter data from array data

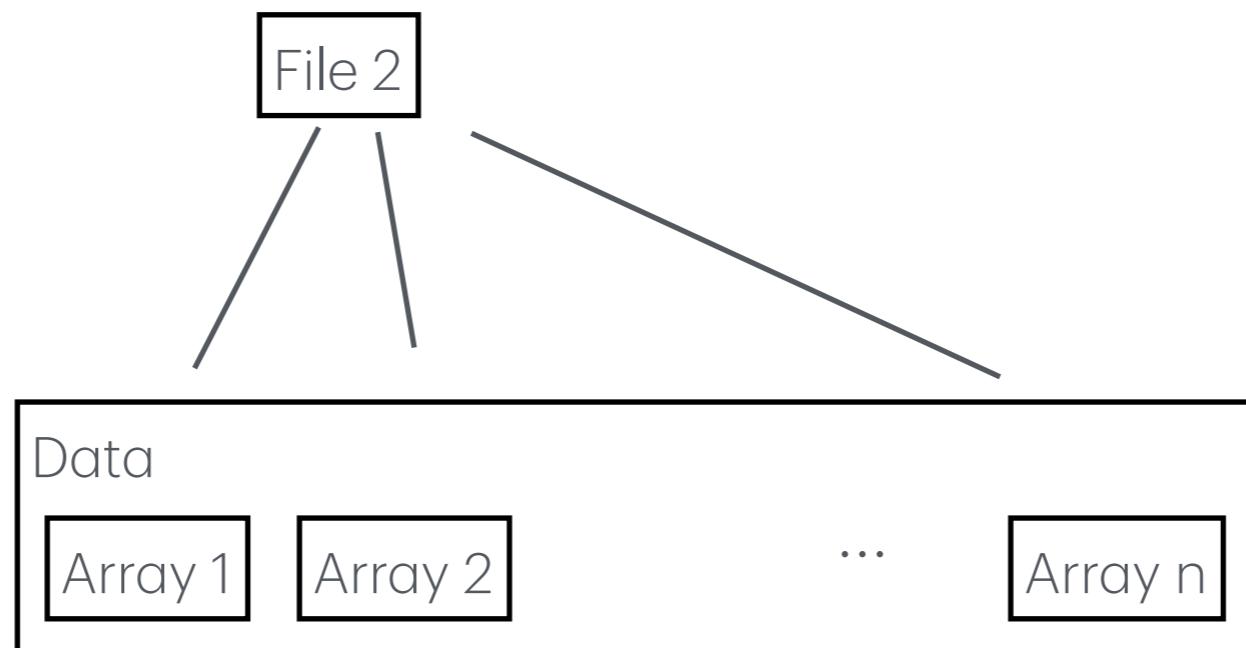
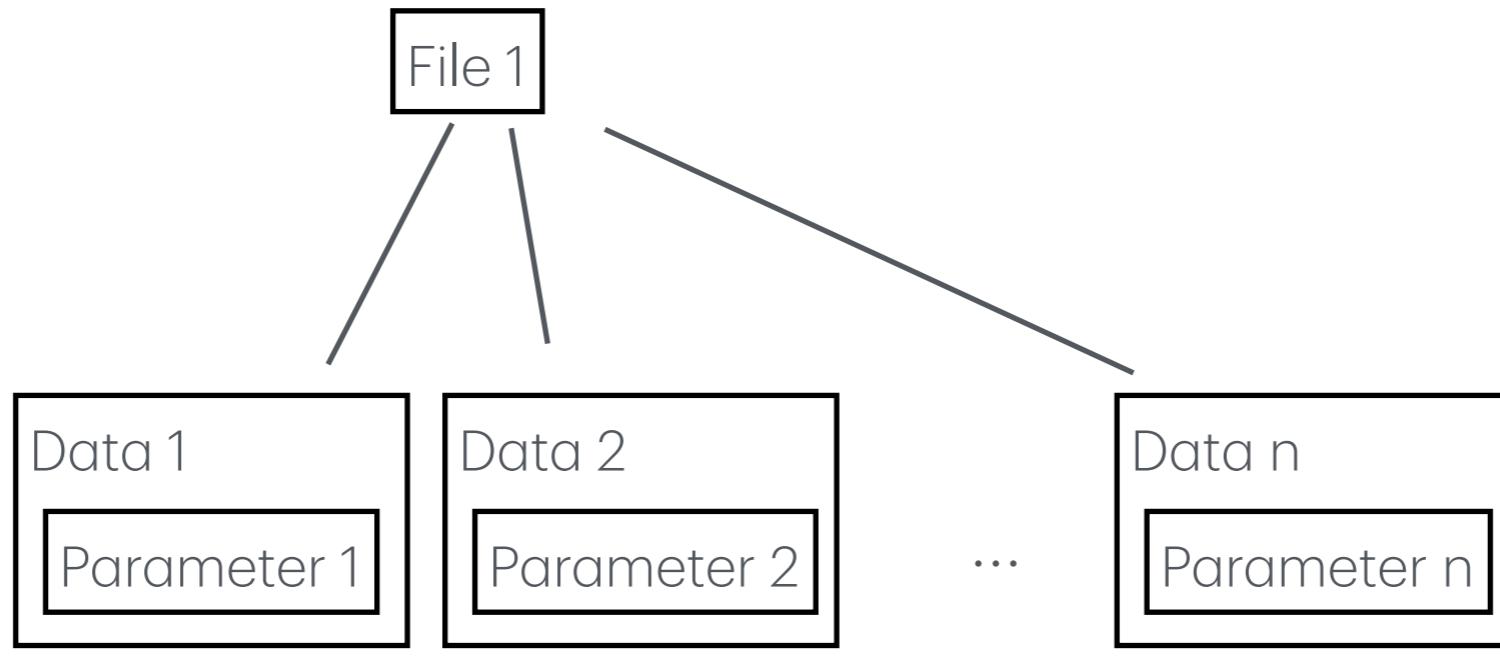
- Before



Storage strategies

Separate parameter data from array data

- After



Storage strategies

Separate parameter data from array data

- Pros
 - More efficient to process
 - Better scalability
 - Easy collaboration
- Cons
 - More complicated to use
 - Not self-contained
 - Potential out of sync

Hands-on

<https://github.com/icui/data-workshop>

Hands-on

Process an earthquake catalog

- Download a database of all earthquake between 1976 and 2020
- Explore storage options for a subset of the database
- Calculate some statistics for all earthquakes
- Calculate some values of each earthquake
- (What if we do these with MPI?)

Hands-on

Part 1: download and extract data

- <https://www.globalcmt.org/CMTsearch.html>
(or search for Global CMT search)
- Click “various ASCII files for direct download”

Catalog in ASCII format

The CMT catalog and Quick CMTs are also available in [various ASCII files for direct download](#).

- Right click link “all events 1976-2020” and copy url

The CMT catalog is available in the ASCII "ndk" format:

- [all events 1976-2020 \(or gzip compressed\)](#).
- Optionally, look at “description of the ndk format”

Note that the "ndk" format is slightly different from the older "dek" format. Please see the [description of the ndk format](#).

Hands-on

Part 1: download and extract data

Copy the url from previous step

```
url='https://www.ideo.columbia.edu/~gcmt/projects/CMT/catalog/jan76_dec20.ndk'
```

Python

```
import urllib.request  
  
with urllib.request.urlopen(url) as response:  
    data = response.read().decode('utf-8')
```

MatLab

```
options = weboptions('ContentType','text');  
data = webread(url, options);
```

Hands-on

Part 1: download and extract data

- Extract the following data for processing
 - Event name
 - Event depth (km)
 - Moment tensor (6 components) with exponent
 - Scalar moment (MO)

| | | | | | | | | |
|-----------|------------|------------|--------|---------|---------|-------|--------------------------------------|---------------------------------|
| MLI | 1976/01/01 | 01:29:39.6 | -28.61 | -177.64 | 59.0 | 6.2 | 0.0 | KERMADEC ISLANDS REGION |
| M010176A | B: | 0 | 0 | 0 | S: | 0 | 0 | M: 12 30 135 CMT: 1 BOXHD: 9.4 |
| CENTROID: | 13.8 | 0.2 | -29.25 | 0.02 | -176.96 | 0.01 | 47.8 0.6 FREE 0-0000000000000000 | |
| 26 | 7.680 | 0.090 | 0.090 | 0.060 | -7.770 | 0.070 | 1.390 0.160 4.520 0.160 -3.260 0.060 | |
| V10 | 8.940 | 75 | 283 | 1.260 | 2 | 19 | -10.190 | 15 110 9.560 202 30 93 18 60 88 |

Hands-on

Part 1: download and extract data

- Try it yourself
 - Go to repository <https://github.com/icui/data-workshop>
 - Finish “Part 1: download and extract data”

Hands-on

Part 1: download and extract data

- Try out these options
 1. Nested dictionary (JSON) `{"M010176A": {"Depth": ..., "M0": ...}}`
 2. Parameter (JSON) + raw array files (.npy or .m)
`{"Rows": ["M010176A", ...], "Columns": ["Depth", "M0", ...]}`
`Array2D[[], []]`
 3. Key-value pairs (HDF5 or ADIOS2)
 - Event name as the key, e.g `{"M010176A": Array([...])}`
 - Data class as the key `{"Depth": Array(...), "M0": Array(...), ...}`

Hands-on

Part 2: process saved data

- Calculate the min, max and median of earthquake depths
- Calculate the moment magnitude of each earthquake and count earthquakes with moment magnitude greater than 8.0

$$M_w = (2/3) * (\log_{10}(M_0) - 16.1)$$

- Find the maximum absolute value among the moment tensor components of all earthquakes
- Save the results

Hands-on

Part 2: process saved data

- Try it yourself
 - Go to repository <https://github.com/icui/data-workshop>
 - Finish “Part 2:process saved data”

Hands-on

Part 2: process saved data

| | Nested | Parameter + array | Key-value (by event) | Key-value (by data class) |
|---------------------|---------------|--------------------------|-----------------------------|----------------------------------|
| Load time | 0.17 | 0.0084 | 5.1 | 0.006 |
| Compute time | 0.15 | 0.0056 | 0.23 | 0.0023 |

Hands-on

Part 2: process saved data

- For a small set of arrays (e.g. 8 arrays in this case), use key value pairs stored as HDF5
- For a large set of arrays, store as separate parameters and array files or ADIOS2

Hands-on

Part 3: accounting for MPI

- Try it yourself
 - Go to repository <https://github.com/icui/data-workshop>
 - Redo Part 2 with MPI using
 - Parameter + array (two files per MPI process)
 - ADIOS2 (single file)

Hands-on

Part 3: accounting for MPI

- Small dataset: ADIOS2 or HDF5
- Large dataset: ADIOS2
- Performance critical: ADIOS2 or raw binary

Quick recap

Have you followed some good practices?

- Save raw data
- Document what is saved
- Document how Mw is calculated
- Choose reasonable file names

100%