

Flywheel Tools

Data Curation and Manipulation on the Flywheel Platform

Tinashe M. Tapera

Introduction

Data curation in the field of neuroimaging is a fundamental task necessary for creating scalable, reproducible science. As the volume, velocity, and variety of these data sets increases and requires greater computational efficiency and expertise, many researchers have opted to have their data hosted by private vendors whose infrastructure may mitigate some of the challenges of storing, organising, and curating massive amounts of data. One such vendor, Flywheel, provides many solutions for these challenges. However, as the scientific landscape expands, researchers ultimately must be able to curate their data themselves on a case-by-case basis to meet novel scientific standards. Additionally, cases inevitably arise in which boilerplate curation tools are unable to correctly curate legacy data, leading to the need for either tedious and error-prone manual curation, or the need for highly skilled programmers and engineers to assist in automated curation. Efforts to formalize the curation of data in neuroimaging include the Brain Imaging Data Structure (BIDS), an open-source specification on how to store and share neuroimaging files. This specification is in continuous development with input from the community, and provides a flexible framework for the development of software pipelines and applications. Flywheel allows users to curate their data using BIDS, but lacks sufficient flexibility for the large variety of use-cases in neuroimaging data warehousing.

This increasing ubiquity and usefulness of BIDS-valid data sets motivated the development of FlywheelTools, a multi-purpose toolkit for interacting with and curating data on the Flywheel platform to meet BIDS standards.

Methods

The FlywheelTools toolkit allows users to follow a reproducible workflow for BIDS curation and audit of their data. This workflow typically includes: inspection of sequences collected during a study; design of a curation schema; implementation of curation schema; curated data inspection and export; and finally audit of data and analyses (see figure 1).

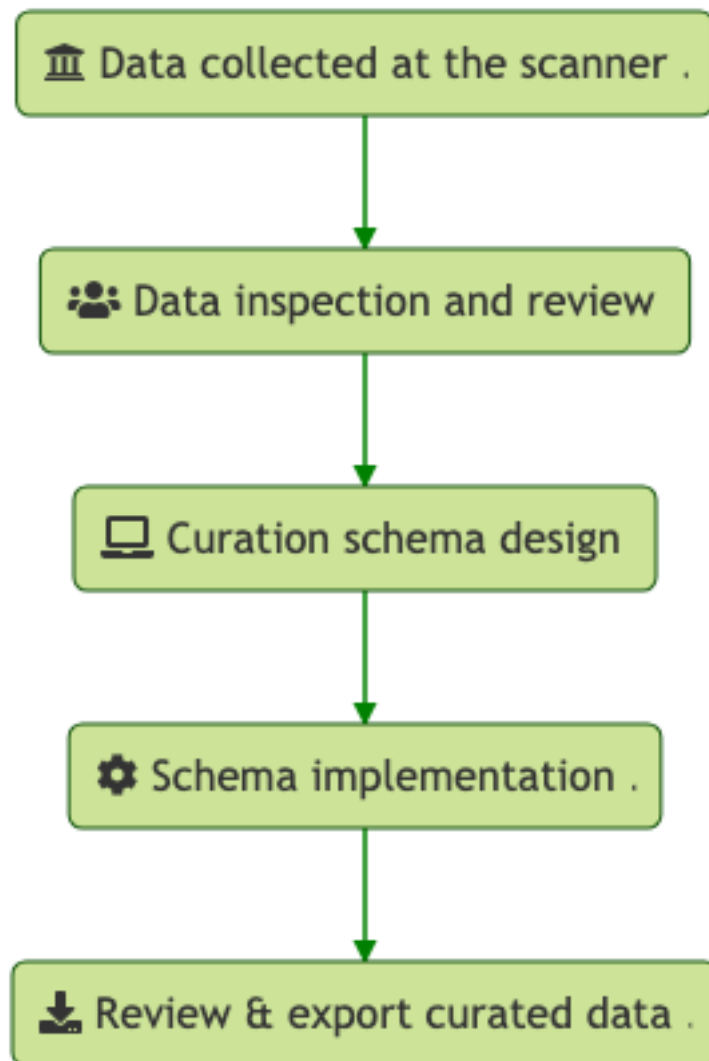


Figure 1: Flywheel Tools Workflow

Programming Languages & Technologies

Flywheel Tools is built primarily in Python 3.6 (Van Rossum and Drake 2009) in order to leverage its highly accessible SDK. Additionally, R 3.4.1 (R Core Team 2019) is used for HTML report generation. For reproducibility and workflow management, Flywheel Tools’ modules are packaged in version-controlled software containers built and managed in Docker (Merkel 2014). Lastly, the Flywheel Tools package relies on users adopting the Brain Imaging Data Structure (BIDS) to curate their data. BIDS has rapidly evolved to become the field standard directory standard in the neuroimaging community for reproducible data organization (Gorgolewski et al. 2016). Importantly, BIDS has a large community of participants contributing to its development and adoption, and proposals for BIDS schema go through rigorous community testing and approval before being added. This leads to a trustworthy common framework for data curation shared amongst researchers. Software developers leverage this ubiquity by creating analysis and processing pipelines that can operate directly on BIDS data sets as inputs, enhancing reproducibility and interoperability on various software platforms.

Flywheel

Flywheel is a data management and analysis platform for research, which lends itself well to neuroimaging. The platform focuses heavily on collaborative and reproducible science. User-facing components of the platform itself are the Web User Interface (UI), the Software Development Kit (SDK), and the Application Programming Interface (API) (IMAGE HERE).

Flywheel Web UI

The web UI is accessible through any modern web browser and is the primary method of interacting with Flywheel data. Through this point-and-click interface, users are able to upload, view, download, and analyse data with ease and simplicity. However, accomplishing tasks with many repetitive steps or a large number of subjects/sessions to iterate over can be tiresome and error-prone. Alongside knowledge of navigating the web UI, many users also make use of the SDK to manipulate and analyse data programmatically.

Flywheel API & SDK

Flywheel’s database utilises MongoDB for data storage and access, meaning that all Flywheel data is represented by hierarchical relationships between document objects. This allows users to create and store complex structures with ease, and query data rapidly (Banker 2011). In order to access this data, Flywheel uses a RESTful Application Programming Interface (REpresentational State Transfer) (Biehl 2016), and hence each document or data object is accessible through a specific URL that a web browser or SDK can access by requesting the data and waiting for a response from the server. The Flywheel Python SDK provides a

powerful interface for inspecting and manipulating data through this API. By standardising this underlying data model into Pythonic Objects, the flywheel SDK is effectively an object relationship mapper (ORM), similar to the popular SQLAlchemy software.

Flywheel Data Model

Objects in Flywheel’s data model follow a specific hierarchical structure (**IMAGE**) — at the top level is the FlyWheel *instance*, a process running that serves the API to users (for example, a neuroimaging center). Within the FlyWheel instance, there are multiple *groups*, which are typically labs or research units that collaborate on one or more *projects*. Each project object can have one or many *subjects* (i.e. participants), and each subject can have one or many *sessions* (i.e. scanning visits). Within a session, there may be one or many *acquisition* objects which represent the scanning sequence collected during a particular scan or examination (e.g. rs-fMRI, DWI, PET), and under each acquisition is *attached* the data file associated with the sequence (e.g. a NIfTI file or DICOM). Note that a file can be attached to any object type, and each object can have metadata associated with it. Hence, a subject object may have de-identified demographic metadata associated to that participant, and the subject may also have a text file attached to it (such as clinical data). A notable exception to the hierarchical structure rule is the analysis object, which behaves in much the same way as others but can be a child object of any other object, allowing researchers to create analyses of entire projects, for example, each with their own associated metadata and files (such as inputs and outputs).

Abstracting this data model in Python results in simple hierarchical objects, each with methods for handling metadata and files, and methods for accomplishing object-specific tasks like traversing the hierarchical structure or running analyses. Flywheel Tools’ modules make use of this data model to accomplish a wide range of tasks.

Flywheel Gears

Flywheel encourages the use of pre-packaged computational workflows, called *gears*. Gears are run by virtual machines/containers using Docker and hence are version-controlled and software/platform agnostic. Gears can accomplish tasks such as data manipulation, pre-processing, analysis, and summarisation. In addition to the multitude of gears available on the platform, users are able to package their own software in a gear and use it for running analysis workflows on their Flywheel data, via the web UI or SDK. the complexity and frequency of the task suggests whether to accomplish a task using the web UI, programmatically using the SDK, or by wrapping it as a workflow into a gear, depends on **IMAGE**. Gears can take existing Flywheel data, such as images or file attachments, as inputs to the workflow, and can be created with clickable configuration options. Once a workflow has completed running, Flywheel collects any files remaining in the container’s pre-defined output directory and attaches them to a resulting

analysis object. The resultant files of a gear (such as an HTML report or tabulated data) can be viewed on the Flywheel UI, downloaded to disk for further data sharing or analysis, or be used as input to a subsequent gear themselves.

Results

Flywheel Tools is implemented using the Flywheel SDK to enable easy inspection, curation, validation, and audit of Flywheel data through a handful of user-friendly gears and command-line interfaces.

The first module of the package is called **fw-heudiconv**, and is largely inspired by the popular HeuDiConv Python package.¹ **fw-heudiconv** is a multi-part toolbox for reproducible curation of neuroimaging data into BIDS on Flywheel. The second module, **flaudit**, is a tool for accomplishing a complete audit of a Flywheel project, giving users a broad overview of the key elements of their data set.

fw-heudiconv

The first tool, **fw-heudiconv**, is a multi-purpose command-line interface and Flywheel gear designed for comprehensive BIDS curation on Flywheel. It is designed to be intuitive, flexible, and reproducible.

Architecture & Design

fw-heudiconv is inspired in large part by the Heuristic Dicom Converter (HeuDiConv) package, and shares much of its design practices. In order to curate data into BIDS, **fw-heudiconv** first considers Dicom data to be the “ground truth” data, and builds its curation approach using data in the Dicoms’ headers. Ultimately, **fw-heudiconv** only has permission to manipulate metadata attached to NIfTI files, in the “Info: BIDS” field, which ensures that curation can be repeated from the stage of Dicom ingress reliably and safely.

fw-heudiconv can be downloaded as a Python command-line interface from the Python Package Index using **pip**, and is available as a point-and-clickable gear on the Flywheel UI. The gear is managed by Docker containerisation, meaning that versioning is reliable and reproducible. There are a number of commands available in **fw-heudiconv**, and each of them starts by querying data from Flywheel. Users can filter their queries, so as to operate on an entire Flywheel project, a subset of subjects, or a subset of sessions. Notably each command has the ability to safely test and evaluate its effects without manipulating metadata on Flywheel or writing data to disk. In particular, there are five commands users can access:

¹<https://github.com/nipy/heudiconv>

1. fw-heudiconv-tabulate The tabulate tool is used to parse and extract DICOM header information in a project (or within a filtered subset of that project) and tabulate this data for the user to examine with ease. By collecting DICOM header information into tabular format, the tabulate tool gives users a comprehensive overview of the different scanning sequences that have been collected in the query, including their sequence parameters. Additionally, users have the option to limit the tabulation to a unique combination of common DICOM header fields, which significantly decreases complexity of the table. The table output by this command can be written to a local disk if run from the command line, or is saved in the output section of a Flywheel gear if run on Flywheel.

2. fw-heudiconv-curate The curate tool is used to curate a dataset on Flywheel into BIDS format. Much like HeuDiConv, curation is done through the use of a heuristic, a Python file which programmatically defines the templates of a range of BIDS valid filenames, and defines the boolean logic that would assign a given scanning sequence to each template. This boolean logic is based on the sequence information users find in the tabulation of sequences, and all fields available in the Dicom header can be used to parse out which template a particular file can be assigned to. Additionally, the curate tool can be used to manipulate BIDS metadata that may need to be added to the dataset. The process of curation only manipulates BIDS metadata of NIfTI files, and hence can be repeated or updated at any time at the user's discretion.

3. fw-heudiconv-export The export tool is used to export a BIDS dataset on Flywheel to disk. The tool is primarily used as a helper tool for other gears and scripts to quickly and easily extract their BIDS data into the workspace of their analysis pipeline, or by individuals who need their BIDS data exported from Flywheel.

4. fw-heudiconv-validate The validate tool is a wrapper around the popular BIDS Validator package **CITE** and is used to check if the applied curation results in a BIDS-valid dataset. After exporting a data set with **fw-heudiconv-export**, the validate tool runs the BIDS validator on the dataset and returns the result and verbose description of the errors and warnings given by the BIDS Validator.

5. fw-heudiconv-clear The clear tool is used to clear BIDS information cleanly and safely from the project or subjects and sessions queried. This can be useful when overwriting by re-curating current BIDS data doesn't fully erase existing data.

Heuristic File

The heuristic file is a Python file used as input to the **fw-heudiconv-curate** command. It instructs **fw-heudiconv** how to programmatically sort and parse

through each acquisition object in Flywheel, and assign it to a BIDS-valid naming template. This is done by checking attributes of a list of `seqInfo` objects — generated from each DICOM’s header information — against user-defined boolean rules. For example, if a T1-weighted image is present in a dataset, the user could define a string with a BIDS-valid naming template for this type of file, such as:

```
t1w = 'subject-{SubjectLabel}_session-{SessionLabel}_T1w.nii.gz'
```

Where the `SubjectLabel` and `SessionLabel` portions are expected to be automatically generated for each subject and session. After a DICOM’s `SeriesDescription` field is added to the `seqInfo`’s `SeriesDescription` attribute, the user can create simple boolean rules to check if the string 'T1w' is in the `SeriesDescription`. If such a rule is met, this acquisition and its NIfTI file will be assigned to the T1-weighted image naming template. The NIfTI file will ultimately have this BIDS naming added to its metadata, and be named correctly when exported to a filesystem.

In addition to setting naming templates, the heuristic file can also be used to hardcode and assign metadata in BIDS. These data are hard-coded into the file object’s metadata on Flywheel, and are assigned by using specially reserved functions and keywords in `fw-heudiconv`. For example, the heuristic file can be used to point fieldmap scans to their intended sequences using a list:

```
IntendedFor = {
    fieldmap1: ['sub-{SubjectLabel}_{SessionLabel}_task-rest_bold.nii.gz']
}
```

By reserving select keywords for functions and metadata, heuristic files become versatile tools for defining and manipulating a wide array of metadata in Flywheel BIDS curation.

Curation Workflow

For most users, the curation workflow follows the sequence detailed above; after ingress of a batch of DICOMs from a scan, Flywheel’s automated utility gears convert the DICOMs to NIfTI files. Users can then begin running `fw-heudiconv-tabulate` to gather the information stored in the DICOM headers necessary for creating a heuristic. Once the tabulation has been completed, the output file can be opened by any program that can read tabular data. Users at this stage can begin creating a heuristic file and running `fw-heudiconv-curate`, using the `--dry-run` flag to test the heuristic changes incrementally with informative logging. When satisfied, users can simply remove the `--dry-run` flag to apply the changes. The user can then use `fw-heudiconv-validate` to run the BIDS validator on the dataset, or start over by removing all BIDS metadata with `fw-heudiconv-clear`.

Additionally, if being run on the Flywheel UI, each of the commands is available as a Flywheel gear. This option can be beneficial for data provenance as all of a gear's commands and inputs, as well as outputs and log files, are stored and attached to each gear run.

flaudit

The second module is a Flywheel project auditor, aptly named **flaudit**. The module is intended to give Flywheel users a broad overview of their entire Flywheel project. Features of this overview include a comprehensive visualization of scanning sequences and their parameters; the corresponding BIDS curated result of each sequence; an enumeration of gear analyses and their runtimes and success rates; and a visualization of the various analysis workflows run on each session **IMAGE**. This information is compiled in a portable HTML report that can be opened in the Flywheel UI or in any web browser.

Architecture & Design

Using similar internal machinery to **fw-heudiconv-tabulate**, **flaudit** loops over existing data in a project and tabulates information about scanning sequences, BIDS metadata, and gear jobs that have been run. These 3 tables are saved internally and then passed as input to an R markdown script that generates a dynamic HTML report. The data are also saved as output for the user to access and analyse at their own discretion.

Discussion

As the neuroimaging community embraces Big Data and the various platforms available for storage and analysis, it is becoming increasingly important for researchers to eschew the *ad hoc* analysis procedures previously run on a single machine or cluster. Instead, cloud-based platforms like Flywheel provide opportunities for more reproducible, reliable, and shareable science. Flywheel Tools provides software that maximizes these opportunities on the Flywheel platform.

Bibliography

- Banker, Kyle. 2011. *MongoDB in Action*. USA: Manning Publications Co.
- Biehl, Matthias. 2016. *RESTful Api Design*. Vol. 3. API-University Press.
- Gorgolewski, Krzysztof J, Tibor Auer, Vince D Calhoun, R Cameron Craddock, Samir Das, Eugene P Duff, Guillaume Flandin, et al. 2016. “The Brain Imaging Data Structure, a Format for Organizing and Describing Outputs of Neuroimaging Experiments.” *Scientific Data* 3 (1): 1–9.
- Merkel, Dirk. 2014. “Docker: Lightweight Linux Containers for Consistent Development and Deployment.” *Linux J*. 2014 (239).
- R Core Team. 2019. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.
- Van Rossum, Guido, and Fred L. Drake. 2009. *Python 3 Reference Manual*. Scotts Valley, CA: CreateSpace.