The Hubble Space Telescope (HST) Advanced Camera for Surveys (ACS) Quicklook Project

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Abstract—The Hubble Space Telescope (HST) Advanced Camera for Surveys (ACS) instrument has been acquiring thousands of astronomical images each year since its installation in 2002 and subsequent restoration in 2009. The ACS Quicklook Project (acsql) provides a means for users to discover and interact with these data via a database-driven web application. This is accomplished via several acsql components: (1) A ~40 TB network file system, which stores all on-orbit ACS data files on disk, (2) a MySQL database, which stores observational metadata in a normalized relational form, allowing users to build custom datasets based on observational parameters, (3) A Python/Flask-based web application, which allows users to view "Quicklook" JPEG images of any publicly-available ACS data along with its metadata, and (4) a Python code library, which provides a platform on which users can build automated instrument calibration and monitoring routines. The acsql project may be extended to support the forthcoming James Webb Space Telescope (JWST) mission, which is scheduled to launch in 2018.

HRC

2004

2006

2008

2010

2012

2014

100k

80k

60k

40k

20k

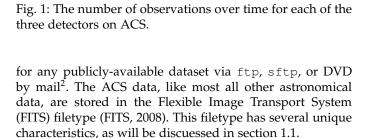
Number of Observations

1 Introduction

The Advanced Camera for Surveys (ACS) is a third-generation imaging instrument on board the Hubble Space Telescope (HST), installed in 2002 during Servicing Mission 3B. It is comprised of three detectors: (1) the Wide Field Camera (WFC), which is designed for wide-field imaging and spectroscopy in visible to near-infrared wavelengths, (2) the High Resolution Channel, which is designed for high resolution near-ultraviolet to near-infrared wavelength images and coronography, and (3) the Solar Blind Channel (SBC), desingned for far-ultraviolet imaging and spectroscopy. ACS expererienced an electronics failure in 2007 that affected the WFC and HRC detectors, until 2009 when astronauts successfully restored the WFC detector during Servicing Mission 4; the HRC still remains unoperational.

Besides these few hiccups, the ACS instrument has been steadily acquiring astronomical images over its 15 on-orbit lifetime. Figure 1 shows an estimates of the number of observations over time for each of the three detectors. To date, there have been nearly 200,000 of observations total. Further information about the ACS instrument including its history, configuration, performance, and scientific capability can be found in the ACS Instrument Handbook (Avila et al., 2017).

ACS data, along with all other data from the other HST instruments past and present (e.g. The Wide Field Camera 3 (WFC3), The Cosmic Origins Spectrography (COS), etc.), are primarily stored and publicly-available in the Barbara A. Mikulski Archive for Space Telescopes (MAST)¹ (Barbara, 2017). Through MAST, users can request and retreive data



The ACS Quicklook Project is a python-based application for discovering, viewing, and querying all publicly-available ACS data. It consists of several subsystems: (1) A filesystem that stores ACS instrument data files and "Quick-

2. Not all HST data are publicly available; most HST data of scientific targets are considered proprietary for up to one calendar year, after which they are publicly released.

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^{1.} named after the U.S. Senator from Maryland who has been a pivitol political driving force behind the manned servicing missions, the Hubble Space Telescope, and the forthcoming James Webb Space Telescope

look" JPEGs in an organized Network File System (NFS), (2) A MySQL database that stores image metadata of each observation, (3) A python/Flask-based web application for interacting with the filesystem and database, and (4) A python code library (named acsql) that contains code for connecting to the database, ingesting new data, logging production code execution, and building/maintaining the database and web application. Each of these subsystems are explained in further detail in the Methodology section of this paper.

This paper aims to outline and detail the ACS Quicklook project as part of the Towson University Computer Science Masters Program Graduate Project. The remaining subsections in this chapter discuss the motivation and use cases for this application, as well as details on the underlying data structure on top of which this project was built. Chapter 2 discusses related work to this project and how the ACS Quicklook project differs from existing similar applications. Chapter 3 details the implementations of each of the ACS Quicklook subsystems. Chapter 4 outlines the results of the project, namely the project deliverables. Lastly, chapters 5 and 6 conclude the paper with a discussion of possible extensions and modifications to the application.

It should be noted that the work that went into this project by the authors was accomplished on behalf of the Space Telescope Science Institute (STScI) located in Baltimore, Maryland. STScI is the home institution for instrument, data, and user support of HST, the forthcoming James Webb Space Telescope (JWST), and MAST. STScI is part of the Association of Universities for Research in Astronomy (AURA).

1.1 Data Structure

The design of the ACS Quicklook application, especially the database, is heavily dependant on the underlying data structure of ACS FITS files. As such, it is important for the reader to understand this data structure and thus the next four sections are dedicated to giving an overview on the subject.

1.1.1 Filenames

Each ACS data file is named in a consistent fashion:

where each <rootname> consists of nine unique alphanumeric characters, and <filetype> is one of several three-character filetype options (discussed in proceeding section 1.1.4). For example, one ACS observation has the rootname j6mf16lhq_raw.fits (Principle Investigator Gary Bernstein, observation date 2016-09-22). Each character in the 9-character rootname has meaning, and is discussed in section 5.2 of the Introduction to the HST Data Handbooks (Smith et al., 2011). The .fits extension at the end of the filename signifies that the file is of FITS format.

Note about rootname caveat.

1.1.2 FITS file structure

Each ACS FITS file consists of several "Extensions", with each extension serving a purpose to describe a particular

TABLE 1: ACS/WFC FITS file extensions

Extension	Purpose	Image Dimensions (pixels)	Data Type
0	Primary header	_	String
1	SCI, Chip 2	(4096, 2048)	Float
2	ERR, Chip 2	(4096, 2048)	Float
3	DQ, Chip 2	(4096, 2048)	Integer
4	SCI, Chip 1	(4096, 2048)	Float
5	ERR, Chip 1	(4096, 2048)	Float
6	DQ, Chip 1	(4096, 2048)	Integer

TABLE 2: ACS/HRC and ACS/SBC FITS file extensions

Extension	Purpose	Image Dimensions (pixels)	Data Type
0	Primary header	-	String
1	SCI	(1024, 1024)	Float
2	ERR	(1024, 1024)	Float
3	DQ	(1024, 1024)	Integer

aspect of the observation. Each extension consists of two parts: (1) an extension "header", which contain key/value pairs describing image metadata (for example, DATE-OBS = '2016-09-22' indicates that the observation date was 2016-09-22) (discussed in the next section), and (2) the extension data, which may be a binary table or, more commonly, a multi-dimensional array of detector pixel values.

The type of extension data can also vary. The most common extension data types are (1) 'science' (SCI), in which the extension data describe a scientific observation, (2) 'error' (ERR), in which the extension data describe the uncertainty in the pixel values of the SCI data, and (3) 'data quality' (DQ), in which the extension data describe the quality of the pixel values for the detector (for example, they may indicate that certain pixels were affected by cosmic rays durring the observation). Typically, for a given file, the 1st extension is the SCI extension, the 2nd extension is the ERR extension, and the 3rd extension is the DQ extension. Furthermore, the 0th extension typically has no extension data and only an extension header that contains metadata that is common to all extensions. This is referred to as the 'Primary Header'.

Tables 1-3 describe the different extensions of ACS FITS files for each of the three ACS detectors. Note that there are two sets of SCI/ERR/DQ extensions for WFC since WFC is comprised of two separate CCD chips.

Over the years, there have been several tools written in various programming languages to read in FITS files and automatically convert their extension data to multidimensional array data types and their extension headers to dictionary data types. For this project, the astropy.fits python library is used extensivly to read and interact with ACS FITS files (Robitaille et al., 2013).

```
bits per data value
                                                      number of data waxes
File may contain standard extensions
Number of standard extensions
                                                   / Number of Standard exacts
/ image is in group format
/ date this file was written (yyyy-mm-dd)
/ name of file
ILENAME= 'j6mf16lhq_raw.fits
ILETYPE= 'SCI '
                                                   ' / name of fi
/ type of data found in data file
                                      / telescope used to acquire data
/ identifier for instrument used to acquire data
2000.0 / equinox of celestial coord. system
                     / DATA DESCRIPTION KEYWORDS
                                                  ' / rootname of the observation set
/ type of exposure identifier
/ instrument designated as prime
 OTNAME= 'j6mf16lhq
AGETYP= 'DARK
RIMESI = 'ACS
                     / TARGET INFORMATION
                ARGNAME= 'DARK
                                         9433 / PEP proposal identifier
' / proposal logsheet line number
' / last name of principal investigator
' / first name of principal investigator
' / middle name / initial of principal investigat
              'Bernstein
                     / EXPOSURE INFORMATION
                                / UT date of start of observation (yyyy-mm-dd)
/ UT time of start of observation (hh:mm:ss)
/ exposure start time (Modified Julian Date)
/ exposure end time (Modified Julian Date)
/ exposure duration (seconds)—calculated
ATE-0BS= '2003-01-27'
      -OBS= '15:20:01'
TART= 5.266663890058E+04
```

Fig. 2: An example header.

1.1.3 FITS file extension headers

As mentioned in the previous section, each FITS extension contains a "header", which contains key/value pairs of metadata associated with the extension data. Such metadata includes various data that describes the astronomical observation (e.g. target name, exposure time, principle investigator name, etc.), telemetry of ACS or HST in general at the time of observation (e.g. temperature of the ACS instrument, orientation of the telescope pointing, position of the telescope relative to Earth, etc.) or the FITS file itself (e.g. the number of extensions, file creation date, etc.). A subsection of an example header is shown in Figure 2.

Extension headers may contain a large number of keyword/value pairs. Some extension headers contain upwards of 300 keywords, while others may contain only \sim 40 keywords.

1.1.4 FITS filetypes for ACS

As discusses in section 1.1.1, each ACS observation may result in several FITS filetypes. Each filetype describes the observation in a different way. The set of available filetypes for a given observation is dependent on the characteristics of the observation, the details of which are beyond the scope of this paper. Also beyond the scope of this paper are the vast details that surround each filetype; each one has a different scientific application that is not important to understanding the ACS Quicklook project. However, to provide at least some context, below we give a brief description of each possible filetype that a given observation may contain:

- raw The raw, uncalibrated data that comes directly from HST
- flt nominally calibrated data
- flc nominally calibrated data plus corrected for Charge Transfer Efficieny (CTE) deficits.
- drz Geometric distortion-corrected data
- drc Geometric distrotion-corrected plus CTE corrected data
- spt Telescope telemetry data
- jit Telescope pointing data
- jif Telescope drifting data
- crj Cosmic ray rejected data
- crc Cosmic ray rejected plus CTE corrected data
- asn Observation association table.

As noted earlier, a given observation may not result in the set of all possible filetypes. For example, the observation j6mf16lhq only has the filetypes raw, flt, jit, jif, and spt.

1.2 Key Metadata

There are several metadata key/value pairs that are particulary important for the ACS Quicklook application, specifically the web application. For some reference, and context, these metadata are briefly described below. Note that the rootname and proposal_type are not metadata from extension headers, but rather are metadata that were explicitly added to the database schema.

APERTURE - The portion of the WFC, HRC, or SBC detector that that was used during an observation. Can either be the entire detector (called a "full-frame image") (e.g. WFC), or a subsection of the detector (called a "subarray") (e.g. WFC1-1K).

DATE-OBS - The date of the observation in the format YYYY-MM-DD, measured in Universal Time (e.g. '2017-08-05').

DEC_TARG - The declination of the target (i.e. the angular distance the target north or south of the celestial equator) (e.g. 41.2842).

DETECTOR - The detector used for the observation. Can either be WFC, HRC, or SBC.

EXPFLAG - Indicates if an observation was interrupted (e.g. INTERRUPTED) or not (e.g. NORMAL).

EXPSTART - The exposure start time of the observation, in units of Modified Julian Date (e.g. 52473.8).

EXPTIME - The exposure duration of the observation, in units of seconds (e.g. 1000.0).

FILTER1 - The selected element from the ACS filter wheel # 1 (e.g. F606W).

FILTER2 - The selected element from the ACS filter wheel # 1 (e.g. F814W).

IMAGETYP - The type of exposure for the observation (e.g. BIAS, EXT, etc.).

OBSTYPE - The type of observation, either IMAGING, SPECTROSCOPIC, CORONOGRAPHIC, or INTERNAL.

proposal_type - The type of proposal that the observation belongs to, such as Calibration (i.e. CAL) or General Observer (i.e. GO).

PROPOSID - The proposal number that the observation belongs to (e.g. 10695).

RA_TARG - The right ascension of the target (i.e. the angular distance of the target east and west on the celestial sphere) (e.g. 49.5375).

rootname - The 8-character unique rootname of the observation (e.g. j5915401).

SUBARRAY - A boolean flag that indicates if the observation is a full-frame APERTURE (i.e. 0) or a subarray APERTURE (i.e. 1).

TARGNAME - The name of the target (e.g. M87, NGC-4536, ANDROMEDA-I, etc.).

TIME-OBS - The time of the start of the observation in the format HH:MM:SS, measured in Universal Time (e.g. 14:21:56).

1.3 Motivation

The motivation for the ACS Quicklook system is driven by several shortcomings of the FITS file structure as well as the current capabilities of MAST from a specific user perspective (inteded users and their use cases are discussed in section 1.2). Some of these shortcomings are described below along with the intended way the ACS Quicklook application will address them.

Data retreival letency: Currently, users who wish to retreive data from the MAST archive must submit a retreival request via the MAST online interface. Once the retreival request is processed (usually automatically unless it is a request of a large number of datasets), the data are either transfered to the user directly via sftp, transfered to a "staging area" in which the user can log into and copy the data via ftp at their leisure, or sent by mail via DVD, depending on which option the user selects. In the case of any one of these options, the time between a download request and the time in which the user has fully retreived the data is a non-significant amount of time. In the fastest scenario of the sftp option, a typical request can take minutes to hours to be completed. The ACS Quicklook system attempts to circumnavigate this retreival process by making the full data products instantly available via readonly access of the filesystem subystem, as well as a subset of the data products (and corresponding metadata) instantly available to view through the web application.

File I/O: Users who
Data redundancy: Something.
Data discovery: Something

1.4 Use Cases

The intended user of ACS Quicklook are ACS instrument scientists, analysts, or scientific users who wish to perform one or more of the following use cases:

1. View

2 RELATED WORK

Topics to discuss:

1. The MAST archive 2. The MAST portal 3. The WFC3/Quicklook project 4. Other Astronomy Institutions 5. How ACS/Quicklook is different

3 METHODOLOGY

In this chapter, we disucss the methods by which we implemented the various subsystems of the ACS Quicklook system. Additionally, we discuss the programming standards and standard workflows that were employed to promote code quality such as readability, maintainability, extensibility, etc; we believe that this aspect of the project is at least equaly important to the system as its individual components.

3.1 Version control

All code associated with this project (including this paper iteself) is version controlled using the git Version Control System (VCS) (git, 2017). The git repository for the project is named acsql. The git repository is also hosted on GitHub, a repository hosting service (GitHub, 2017), and is publicly available at http://github.com/spacetelescope/acsql/.

Several feature branches of the code were created throughout the building of this project such that the master branch (which is considered the "production" branch) always contained operational code (while the code in the branches may contain unfinished implementations). Such branches include create-database (for implementation of the database schema), add-logging (for implementation of system logging), build-ingest (for implementation of data ingestion software), and web-application (for implementation of the web application). For each merge of a feature branch, a tag and release was created for the master branch to be saved in the repository. These releases are available at https://github.com/spacetelescope/acsql/releases.

Additionally, using GitHub allowed for issue tracking of bugs, features, and potential enhancements to the code repository. Current open issues of the repository can be found at https://github.com/spacetelescope/acsql/issues.

```
def get_proposal_type(proposid):
    """Return the ``proposal_type`` for the given ``proposid``.

The ``proposal_type`` is the type of proposal (e.g. ``CAL``,
    '`GO``, etc.). The ``proposal_type`` is scraped from the MAST
    proposal status webpage for the given ``proposid``. If the
    '`proposal_type`` cannot be determined, a ``None`` value is returned.

Parameters
------
proposid : str
    The proposal ID (e.g. ``12345``).

Returns
------
proposal_type : int or None
    The proposal type (e.g. ``CAL``).
```

Fig. 3: An example of the PEP257 and numpydoc docstring conventions, using the get_proposal_type function from acsql.ingest.ingest.

3.2 Programming and Documentation Standards

All code contained within this project was written to adhere to specific standards and conventions, namely (1) the PEP8 Style Guide for python code (van Rossum, 2001), (2) The PEP257 python guide for module and function docstring conventions (Goodger, 2001), and (3) the numpydocs documentation standard (NumPy Documentation, 2017). More details on each of these standards and conventions are given below.

The PEP8 Style Guide for python code (abbreviated for 'Python Enhancement Proposal #8') documents python coding conventions including variable naming, spacing, line length, module layout, function layout, comments, and design patterns. Only in specific cases were these conventions not followed, such as using a single line of code, even if it exceeded the recommended 80 characters, to allow for greater readability. By following these conventions, the style of the acsql code is constistent amongst each module and attempts to reflect the style of industry-grade python code.

The PEP257 guide for docstring conventions describes standard conventions used for function and module docstrings (i.e. the API documentation found in block comments at the beginning of modules or immediately after function declarations). Like PEP8, following these conventions ensure consistency amongst the acsql code documentation. Furthermore, the numpydocs documentation convention provides some additional details on top of the PEP257 conventions and is used in many python packages including the numpy (numerical python) and scipy (scientific python) packages (van der Walt et al., 2011). Figure N shows an example of these conventions, taken from the ascql.ingest.ingest.get_proposal_type function.

Another benefit to using PEP257 and numpydoc docstring conventions is that API documentation creation tools such as sphinx (Brandi et al., 2007) or epydoc (Loper, 2004) can automatically convert the docs into other output formats such as HTML and PDF. For this project, we use sphinx to convert API documentation to HTML, and host the webpages online using the readthedocs, which is an open-

```
ingest.ingest.get_proposal_type(proposid)

Return the proposal_type for the given proposid.

The proposal_type is the type of proposal (e.g. CAL , 60 , etc.). The proposal_type is scraped from the MAST proposal status webpage for the given proposid. If the proposal_type cannot be determined, a None value is returned.

Parameters: proposid: str

The proposal ID (e.g. 12345 ).

Returns: proposal_type: int or None

The proposal type (e.g. CAL ).
```

Fig. 4: The readthedocs documentation for the acsql example function seen in Figure N.

```
filesystem/
jcp0/
jcp001kwq/
jcp001kwq_flc.fits
jcp001kwq_flt.fits
jcp001kwq_raw.fits
jcp001kwd_spt.fits
jcp001kwj_jit.fits
jcp001010/
jcp001010_drs.fits
jcp001010_drc.fits
jcp001010_jif.fits
jcp001010_jif.fits
jcp001010_jif.fits
jcp001010_jif.fits
jcp001010_jif.fits
jcp001010_jif.fits
jcp0014tyq/
...
jcp001ktq/
...
jcp7/
...
```

Fig. 5: A representation of the directory structure within the acsql filesystem, using a few observations as an example.

source, community supported tool for hosting and browing documentation (Read the Docs, 2017). The documentation for acsql is hosted at http://acsql.readthedocs.io/. The output documentation as seen on readthedocs for the example function in figure N is provided in Figure N.

3.3 Filesystem: Archive of ACS data

The acsql filesystem is a Network File System (NFS) that stores all on-orbit ACS data on disk in an organized set of directories and subdirectories hosted at STScI. Figure N shows an example of this directory structure: The parent directory is the first four characters of the 9-character rootname, which maps directly to an individual PROPOSID. The subdirectories of the parent directories are named after the full 9-character rootname such that each parent directory contains the rootname subdirectories that were observed for that particular PROPOSID. Each rootname subdirectory contains every available filetype (as described in Section 1.1.4) for the particular observation is stored.

Figure N shows how the total size of the filesystem has evolved over the lifetime of the mission; currently, the filesystem occupies \sim 40 TB of storage space. Note that the file sizes across the detectors and across the various filetypes

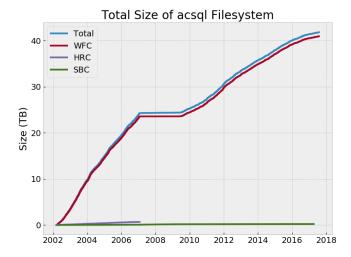


Fig. 6: The size of the acsql filesystem as a function of observation date.

may vary depending on the nature of the particular obseration (for example, full-frame observations result in larger file sizes than subarray observations, calibrated filetypes have larger file sizes than un-calibrated filetypes, etc.).

3.4 Filesystem: Archive of JPEGs and Thumbnails

In addition to the ACS data products described in the last section, the <code>acsql</code> filesystem also stores "Quicklook" <code>JPEG</code> and <code>thumbnail</code> images of each <code>RAW</code>, <code>FLT</code>, and <code>FLC</code> filetype (when applicable) in an organized directory structure. These images are used by the <code>acsql</code> web application to allow users to quickly and easily view the contents of the data without having to pysically open the corresponding <code>.fits</code> files.

The JPEGimages are are generated by taking the two-dimensional data from the SCI extension(s), sigma-clipping the top and bottom 1% of the values (as to avoid large outlier values and enhance the scaling of the image), and saving the data to a a JPEG format. The thumbnail images are created by simply resizing the corresponding JPEG into a 128x128 pixel image and saving to a .thumb extension; the purpose of these thumbnail images are to be able to view many of them on a single webpage in the acsql web application. An example of a JPEG image and its corresponding thumbail is shown in Figure N.

Unlike the ACS data products portion of the filesystem, the JPEG and thumbnail portions of the filesystem are organized based on the 5-digit PROPOSID of the corresponding observation instead of the first four characters of the rootname. This design was chosen as a means to simplify the design of the web application; users often which to view data based on the 5-digit PROPOSID and less often on the details of the rootname. An example of this sturcture is shown in Figure N. Note that the thumbail filesystem only contains thumbnails created from FLT filetypes, since thumbnails are only inteded for navigation and quickviewing.

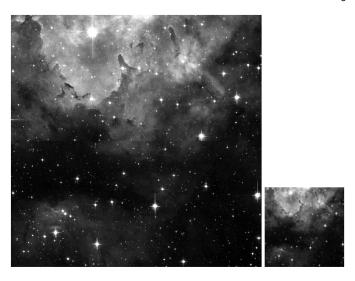


Fig. 7: An example of a JPEG image (left) and its corresponding thumbnail image using example dataset jcs718koq.

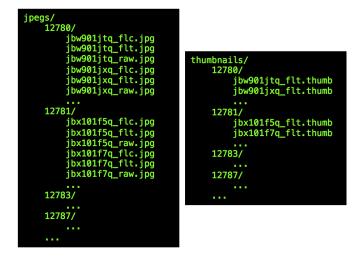


Fig. 8: A representation of the directory structure for the JPEG (left) and thumbail (right) portion of the acsql filesystem, using a few observations as an example.

3.5 Database: Relational Schema

Another major component of the acsql project is a relational database that stores all FITS header key/value pairs for each ACS filetype and FITS file extension across all on-orbit ACS observations. Such a database allows users to perform relational queries for any observational metadata.

To accomplish this, we implemented the relational schema shown in Figure N. The acsql database contains 111 tables in total: one master table which contains basic information about each rootname that is important for the acsql database in general, one datasets table which indicate which filetypes are available for a particular rootname, and 109 'header' tables which stores the header key/value pairs, one for each detector/filetype/extension combination (e.g. wfc_raw_0). Each of these tables are described in detail below.

The master table contains information that is particularly useful for maintaining and using the acsql database.

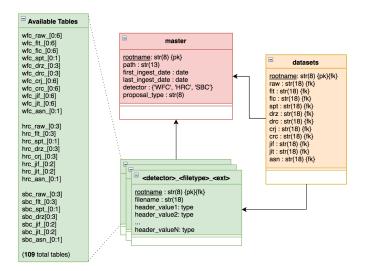


Fig. 9: The relational database schema for the acsql database.

Its primary key is the first 8 characters of the 9-character rootname for the partiuclar observation (recall from section 1.1.1 that only the first 8 characters of a rootname are actually unique). The path column contains the location of the observation in the acsql filesystem. The first_ingest_date and last_ingest_date contains the date in which the observation was first inserted into the database and the date in which the observation was most recenly updated in the database, respectively. The last_ingest_date allows the database maintainer to determine when data in the database may become outdated and require re-ingestion.

The datasets table lists which filetypes are available for each observation. If a particular filetype is available for the given rootname, the value for the appropriate column in the table is the full <rootname>_<filetype>.fits filename (for example, the raw column contains the value jcs718koq_raw.fits for rootname jcs718ko). If a particular filetype is not available, the value of the column is NULL. This table allows a user to determine which header tables are queryable for a given rootname. The rootname in the datasets table acts as both a primary key for the table as well as a foreign key that maps to the rootname of the master table.

The remaining 109 tables were designed to be in direct correspondance with the header metadata key/value pairs found in observations files; each column is named in the same manner as the header keys, with the value of that column reflecting the header value. There is one table for each detector, filetype, and extension combination; collectively, these are referred to as the 'header' tables. Like with the datasets table, the rootname column serves as a primary key for the header tables as well as a foreign key that maps to the rootname of the master table.

3.6 Database: MySQL + SQLAlchemy

The acsql database is stored on a MySQL server (Version 5.6) (Oracle, 2017) that is hosted at STScI. The database schema was implimented using SQLAlchemy, which is

an open-source SQL toolkit and Object Relational Mapper (ORM) for python (Bayer, 2006). As an ORM, SQLAlchemy enables python classes to be easily translated to SQL-based database tables, and vice versa. Additionall, SQLAlchemy provides python methods for connecting to a SQL-based database and performing typical SQL tasks such as inserts, updates, and queries.

There are several kev functions and classes that were used to construct the acsql database (all of which can be found in acsql.database.database_interface.py module). One such function is the load_connection, as shown in Figure N. This function creates three SQLAlchemy objects that are used to establish a connection with the acsql database: engine, base, and session, each described below.

The engine object contains the Python Database API Specification (also known as DBAPI), which provides a low-level API for python-specific, commonly-used database tasks (Lemburg, 2017). It is created from sqlalchemy.create_engine method, which requires a user-supplied connection_string. The connection_string is a string that contains information about the type of database, the specific database dialect being used, and the user credentials (e.g. username, password, port number, and host server name). In the case of the acsql, this connection string takes the form of 'mysql+pymysql://username:password@host:port/acsql'. The connection string is imported from a user supplied config file within the acsql library (as will be

The base object serves as a base class for declarative class definitions (i.e. the classes that are used to define the database tables). It is created from the sqlalchemy.ext.declarative.declatative_base method. Perhaps most importantly, the base object contains methods for creating and dropping tables from the class definitions (e.g. base.metadata.create_all() and base.metadata.drop_all(), respectively).

discussed in section 3.9).

The session object provides a primary usage interface for database operations, and is created via the sqlalchemy.sessionmaker method, which takes as a parameter the engine object. The methods of the session object are primarily used to query the database (i.e. session.query()) as well as committing inserts or updates (i.e. session.commit()).

The master and datasets tables were implemented via explicit class definitions in database_interface, and are shown in Figures N and N, respectively. Each table column is defined using the sqlalchemy.Column object, which is a class that can be initialized with the datatype that will be stored in the column (e.g. String, Float, Integer, etc.) as well as parameters that set SQL-like constraints and parameters on the column values. These include, but are not limited to, primary keys (e.g. the primary_key=True parameter in the master.rootname column), foreign key constrains (e.g. the ForeignKey constraint in the datasets.rootname column), uniqueness constraints (e.g. the unique=True parameters in the master.path column), and NULL constraints (e.g. the nullable=False parameter in the master.first_ingest_date column).

```
def load connection(connection string):
    """Return ``session``, ``base``, and ``engine`` objects for
    connecting to the ``acsql`` database.
    Create an ``engine`` using an given ``connection_string``. Create a
     `base`` class and ``session`` class from the ``engine``. Create an
    instance of the ``session`` class. Return the ``session``,
    ``base``, and ``engine`` instances.
    Parameters
    connection string : str
        The connection string to connect to the ``acsql`` database. The \ensuremath{\mathsf{C}}
        connection string should take the form:
        ``dialect+driver://username:password@host:port/database``
    Returns
    session : sesson object
       Provides a holding zone for all objects loaded or associated
       with the database.
    base : base object
        Provides a base class for declarative class definitions.
    engine : engine object
       Provides a source of database connectivity and behavior.
    engine = create_engine(connection_string, echo=False, pool_timeout=100000)
    base = declarative base(engine)
    Session = sessionmaker(bind=engine)
    session = Session()
    return session, base, engine
```

Fig. 10: The load_conenction function, which is used to build a connection to the acsql database

Fig. 11: The class definition for constructing the master table via SQLAlchemy.

SQLAlchemy determines the name of the table via the __tablename__ attribute, and determines the name of the columns by the name of the variable used to initialize each Column object.

Since there are 109 header tables, some of which have hundreds of columns, it is not practical to construct a class definition for each table in a similar manner to that of the master and datasets table. Instead, these class definitions were implemented via the database_interface.orm_factory function, which is a factory function that creates and returns a class definition

```
class Datasets(base):
    """ORM for the datasets table."""
    def __init__(self, data_dict):
        self.__dict__.update(data_dict)
     _tablename__ = 'datasets'
    rootname = Column(String(8), ForeignKey('master.rootname'),
                      primary_key=True, index=True, nullable=False)
    raw = Column(String(18), nullable=True)
    flt = Column(String(18), nullable=True)
    flc = Column(String(18), nullable=True)
    spt = Column(String(18), nullable=True)
    drz = Column(String(18), nullable=True)
    drc = Column(String(18), nullable=True)
    crj = Column(String(18), nullable=True)
    crc = Column(String(18), nullable=True)
    jif = Column(String(18), nullable=True)
    jit = Column(String(18), nullable=True)
    asn = Column(String(18), nullable=True)
```

Fig. 12: The class definition for constructing the datasets table via SQLAlchemy.

```
def orm_factory(class_name):
   """Create a SQLAlchemy ORM Classes with the given ``class_name``.
   class name : str
       The name of the class to be created
   Returns
   class : obi
      The SQLAlchemy ORM
   data_dict = {}
   data_dict['__tablename__'] = class_name.lower()
   data_dict['rootname'] = Column(String(8), ForeignKey('master.rootname'),
                                  primary_key=True, index=True,
                                  nullable=False)
   data_dict['filename'] = Column(String(18), nullable=False, unique=True)
   data_dict = define_columns(data_dict, class_name)
   data_dict['__table_args__'] = {'mysql_row_format': 'DYNAMIC'}
   return type(class_name.upper(), (base,), data_dict)
```

Fig. 13: The orm_factory function, used to define class definitions for header tables.

for each header table, based on the given class_name that reflects the detector/filetype/extension combination (e.g. wfc_raw_0). The orm_factory function is shown in Figure N. Similar to the Master and Datasets classes, some of the columns in the orm_factory function are explicitly defined via the SQLAlchemy Column class. However, the columns that correspond to header key/value pairs are defined in a separate function named define_columns, shown in Figure N.

The purpose of the define_columns function is to define SQAlchemy Column objects for each header keyword in the headers of the particular detector/filetype/extension combination

```
def define columns(data dict, class name):
    """Dynamically define the class attributes for the ORM
   Parameters
   data_dict : dict
      A dictionary containing the ORM definitions
   class name : str
      The name of the class/ORM.
   data dict : dict
      A dictionary containing the ORM definitions, now with header
      definitions added.
   special_keywords = ['RULEFILE', 'FWERROR', 'FW2ERROR', 'PROPTTL1',
                       'TARDESCR', 'OUALCOM2']
   with open(os.path.join(os.path.split(__file__)[0], 'table_definitions',
                         class_name.lower() + '.txt'), 'r') as f:
       data = f.readlines()
   keywords = [item.strip().split(', ') for item in data]
   for keyword in keywords:
       if keyword[0] in special keywords:
           data_dict[keyword[0].lower()] = get_special_column(keyword[0])
       elif keyword[1] == 'Integer':
           data dict[kevword[0].lower()] = Column(Integer())
       elif keyword[1] == 'String':
           data_dict[keyword[0].lower()] = Column(String(50))
       elif keyword[1] == 'Float':
           data_dict[keyword[0].lower()] = Column(Float(precision=32))
       elif keyword[1] == 'Decimal':
           data_dict[keyword[0].lower()] = Column(Float(precision='13,8'))
       elif keyword[1] == 'Date':
           data_dict[keyword[0].lower()] = Column(Date())
       elif keyword[1] == 'Time':
           data_dict[keyword[0].lower()] = Column(Time())
       elif keyword[1] == 'DateTime':
          data_dict[keyword[0].lower()] = Column(DateTime)
       elif keyword[1] == 'Bool':
           data dict[keyword[0].lower()] = Column(Boolean)
           raise ValueError('unrecognized header keyword type: {}:{}'.format(
               keyword[0], keyword[1]))
       if 'aperture' in data dict:
           data_dict['aperture'] = Column(String(50), index=True)
```

Fig. 14: The define_columns function, used to define columns used in the header tables.

(provided in the given class_name parameter). This is accomplished by reading in a text file (named <class_name>.txt that contains the header keywords and their datatype (one per line) for the given class_name. An portion of an example text file is shown in Figure N.

Furthermore, the 109 text files used to define the header table columns are also generated in an automated fashion via the acsql.database.make_tabledefs.py module. This module uses a set of example FITS files to scrape its header contents, determine all of the header keywords and their datatypes, and write the results to a text file. Similarly, the acsql.database.update_tabledefs.py is used to add new header keywords by comparing the header contents of a given FITS file and the existing column

```
DETECTOR, String
NEXTEND, Integer
EXTEND, Bool
SIMPLE, Bool
NAXIS, Integer
LINENUM, String
GROUPS, Bool
DATE, String
EQUINOX, Float
INSTRUME, String
PROPOSID, Integer
ASN_ID, String
ASN PROD, Bool
ASN_STAT, String
DEC TARG, Float
FILETYPE, String
ASN_TAB, String
PRIMESI, String
RA_TARG, Float
TARGNAME, String
TELESCOP, String
PR_INV_F, String
BITPIX, Integer
PR INV_M, String
PR_INV_L, String
```

Fig. 15: The contents of an example text file used to define the columns of a header table in the define_columns function. The example table used here is the wfc_asn_0 table.

```
WFC raw 0 = orm factory('WFC raw 0')
WFC_raw_1 = orm_factory('WFC_raw_1')
WFC_raw_2 = orm_factory('WFC_raw_2')
WFC_raw_3 = orm_factory('WFC_raw_3')
WFC raw 4 = orm factory('WFC raw 4')
WFC_raw_5 = orm_factory('WFC_raw_5')
WFC_raw_6 = orm_factory('WFC_raw_6')
WFC flt 0 = orm factory('WFC flt 0')
WFC_flt_1 = orm_factory('WFC_flt_1')
WFC_flt_2 = orm_factory('WFC_flt_2')
WFC_flt_3 = orm_factory('WFC_flt_3')
WFC flt 4 = orm factory('WFC flt 4')
WFC_flt_5 = orm_factory('WFC_flt_5')
WFC_flt_6 = orm_factory('WFC_flt_6')
WFC flc 0 = orm factory('WFC flc 0')
WFC_flc_1 = orm_factory('WFC_flc_1')
WFC_flc_2 = orm_factory('WFC_flc_2')
WFC_flc_3 = orm_factory('WFC_flc_3')
WFC_flc_4 = orm_factory('WFC_flc_4')
WFC_flc_5 = orm_factory('WFC_flc_5')
WFC_flc_6 = orm_factory('WFC_flc_6')
```

Fig. 16: An example of how the orm_factory function is called to create class defintions for the header tables.

definition text files³.

With the implementation of the orm_factory and define_columns function, it is then trivial to create class definitions for each of the 109 header tables. An example of this is shown in Figure N, where the several of the WFC header tables are defined.

With the master, datasets, and each of the 109 header tables defined in the database_interface module, creating the database tables on the MySQL server is accom-

3. New header keywords are occaisonally introduced to ACS data proceeding updates to its calibration software

plished by executing the $base.metadata.create_all()$ method.

3.7 Data ingestion software: Algorithm

Another critical component of the ACS Quicklook system are the modules that are used to ingest data into the acsql database and to create the "Quicklook" JPEGs and thumbails. By the term 'ingest', we refer to the following algorithm:

- 1. Identify newly available public ACS data in the filesystem: This is accomplished by comparing the list of rootnames in the filesystem with the list of rootnames in the master table of the acsql database. Any rootnames that exist in the filesystem but not in the database are considered new rootnames to be ingested.
- 2. Loop over each rootname (in a parallelized manner): The ingestion software (i.e. the acsql.ingest.ingest module), takes as input a single rootname, such that if there are multiple rootnames to be ingested, the calls to the ingestion module can be parallelized over many CPUs. The ingestion of one rootname does not depend on the ingestion of another, nor is the order of which files are ingested important. Please note that steps 3 through N are written from the perspective that a single rootname is being ingested (i.e. inside of the loop.)
- 3. Update the master table with information about the rootname: At this point, the master table can be updated with metadata pertaining to the rootname. A generic insert_or_update function was written (available in the acsql.utils.utils module) to determine if an entry should be inserted (in the case of first-time ingestion) or updated (in the case of re-ingestion). This function uses various sqlalchemy methods and the class definitions described in section 3.6 to perform the insert or update operation. The insert_or_update function is shown in Figure N.
- 4. Loop over the available filetypes for the given rootname: The available filetypes are determined by traversing down a level in the tree structure of the filesystem and identifying which files are present. Once determined, the ingestion algorithm processes each <rootname>_<filetype>.fits file individually. Please note that steps 5 through N are written from the perspective that a single file is being ingested (i.e. inside of the next nested loop).
- 5. Create a python dictionary with metadata about the file: To reduce the amount of variables being passed around to various functions, a data container in the form of a python dictionary data type is created to hold metadata needed by the remainder of the ingestion process. We refer to this data container as the file_dict. The file_dict contains metadata such as the absolute path of the file in the filesystem, the filetype, the available FITS file extensions of the file, and the absolute paths to which the "Quicklook" JPEGs and Thumbnails will be written.
- 6. For each FITS file exentension, extract the header information and update the appropriate header table in the acsql database: The header information is read into a python dictionary via the astropy.io.fits module. Besides some minor fixes for a few corner cases (such as

```
def insert or update(table, data dict):
    """Insert or update a record in the given ``table`` with the data
   in the ''data dict''.
   A record is inserted if the primary key of the record does not
    already exist in the ``table``. A record is updated if it does
    already exist.
    Parameters
    table : str
       The name of the table to insert/update into.
    data dict : dict
       A dictionary containing the data to insert/update.
    table_obj = getattr(acsql.database.database_interface, table)
    session, base, engine = acsql.database.database_interface.\
        load_connection(SETTINGS['connection_string'])
    # Check to see if a record exists for the rootname
    query = session.query(table_obj)\
        .filter(getattr(table_obj, 'rootname') == data_dict['rootname'])
    query_count = query.count()
    # If there are no results, then perform an insert
    if not query_count:
       tab = Table(table.lower(), base.metadata, autoload=True)
       insert_obj = tab.insert()
            insert_obj.execute(data_dict)
        except (DataError, IntegrityError, InternalError) as e:
            logging.warning('\tUnable to insert {} into {}: {}'.format(
                            data_dict['rootname'], table, e))
    else:
        query.update(data_dict)
   session.commit()
   session.close()
   engine.dispose()
```

Fig. 17: The insert_or_update function from the acsql.utils.utils module, used at various times during the data ingestion process to determine if an entry should be inserted or updated in the acsql database.

converting hypens in header keys to underscores as to avoid python errors), it is rather trivial to perform and insert or update operation via the insert_or_update function (see Figure N).

- 7. Update the datasets table for the given filetype: At this point, an entry in the datasets table is either inserted if it is the first filetype for the rootname being ingested, or updated if a filetype under the same rootname had already been ingested.
- 8. If the filetype is either raw, flt, or flc, then create a "Quicklook" JPEG image: JPEGs are produced only for raw, flt, and flc filetypes, since it are these filetypes that contain actual two-dimensional image data. The image data are read into multidimensional numpy array data types via the astropy.io.fits module. The data are then rescaled as to avoid an undesirable image stretch caused by extremely high or low-valued pixels, and

saved to a .jpg format. The JPEGs are saved to the JPEG portion of the acsql filesystem (described in section 3.4).

9. If the filetype is flt, then create a "Quicklook" Thumbnail image: Thumbnail images are only produced for flt filetypes since they are only meant to be viewed as a means to discover the larger JPEG images via the acsql web application. Thumbnails are generated by simply opening up the corresponding flt JPEG and resizing it to 128x128 pixels. The Thumbnailss are saved to the Thumbnail portion of the acsql filesystem (described in section 3.4).

This workflow is encapsulated within several modules across the acsql.scripts and acsql.ingest subpackages, as will be desribed in section 3.9. These modules are intended to be executed daily (as an automatically-spawned process) as to keep the ACS Quicklook system up-to-date on any public data as it becomes available.

3.8 Data ingestion software: logging

Since the data ingestion software is intended to be executed by an automatic process and not by a human, we implemented a system by which the status of the ingestion process can be logged to an output text file and analyzed at a later time. Such log files can be used to assess if there were any issues with the ingestion process, such as if a new header keyword has appeared (requiring an update to the appropriate header table in the database). An example log file showing the ingestion of a single rootname (j8zh21xv) is provided in Figure N.

When the ingestion module gets executed, an empty log file is created with the filename <module_name>_<timestamp>.log, where <module_name> is the name of the ingestion module (in production, this is ingest_production.py, as will be discussed in section 3.9), and <timestamp> is the current time in the format YYYY-MM-DD-HH-MM. The naming convention of the log file allows system maintainers to determine which log file corresponds to which ingestion run.

Next, the python logging module is used to configure the format of the log statements. It does this by (1) setting the default logging level to INFO (meaning that, unless otherwise specified, each call to logging by the ingestion module will result in an INFO statement.), (2) setting the timestamp format to YYYY-MM-DD HH:MM:SS, and (3) setting the logging message format to <timestamp> <level>: <message>.

With the logging settings configured, any call to the logging module within the ingestion software results in a log statement. For example, the code logging.info('Gathering files to ingest') results in a timestamped log message, e.g. 08/15/2017 11:05:26 INFO: Gathering files to ingest (as shown in Figure N.)

Calls to the logging module are strategically placed within the ingestion software to provide enough context to the status of the ingestion without cluttering the log file with too much detail. In most cases, logging statments only occur after a change of state to the system (i.e. an updated database table, the creation of a JPEG or Thumbnail.)

3.9 Web Application

The front-end of the acsql system is the web application.

3.10 acsql Package

All code associated with the acsql project is contained within a single git repository (also named acsql), which we refer to as the "acsql Library", or "acsql package". The package layout is shown below:

```
acsql/
   LICENSE
   README.md
   MANIFEST.in
   setup.py
   paper/
   presentation/
       . . .
   docs/
       Makefile
        requirements.rst
        source/
          conf.py
          database.rst
          index.rst
          ingest.rst
          scripts.rst
          utils.rst
          website.rst
    acsql/
         _init__.py
        database/
            __init__.py
            database_interface.py
           make_tabledefs.py
           queries.py
            reset_database.py
           table_definitions/
                *.txt
           update_tabledefs.py
        ingest/
            __init__.py
            ingest.py
            make_file_dict.py
            make_jpeg.py
           make_thumbnail.py
        scripts/
             _init__.py
            ingest_production.py
        utils/
            __init__.py
            config.yaml
           utils.py
        website/
            __init__.py
            acsql_webapp.py
            data_containers.py
            form_options.py
            query_form.py
            query_lib.py
            static/
                css/
                    *.CSS
                img/
                    jpegs
                    thumbnails
                js/
                    *.js
            templates/
                *.html
```

We now provide a brief description of each package component:

```
08/15/2017 11:05:26 INFO: User: bourque
08/15/2017 11:05:26 INFO: User: bourque
08/15/2017 11:05:26 INFO: Python Version: 3.5.2 [Continuum Analytics, Inc.] (default, Jul 2 2016, 17:53:06) [GCC 4.4.7 20120313 (Red Hat 4.4.7-1)]
08/15/2017 11:05:26 INFO: Python Path: /bourque/envs/anaconda3/envs/astroconda3/bin/python
08/15/2017 11:05:26 INFO: Numpy Path: /bourque/envs/anaconda3/envs/astroconda3/lib/python3.5/site-packages/numpy
08/15/2017 11:05:26 INFO: Numpy Path: /bourque/envs/anaconda3/envs/astroconda3/lib/python3.5/site-packages/numpy
08/15/2017 11:05:26 INFO: Astropy Version: 1.2.1
08/15/2017 11:05:26 INFO: Astropy Path: /bourque/envs/anaconda3/envs/astroconda3/lib/python3.5/site-packages/astropy
08/15/2017 11:05:26 INFO: Astropy Path: /bourque/envs/anaconda3/envs/astroconda3/lib/python3.5/site-packages/soltalchemy-1.1.4-py3.5-linux-x86_64.egg/sqlalchemy
08/15/2017 11:05:26 INFO: SQlalchemy Path: /bourque/envs/anaconda3/envs/astroconda3/lib/python3.5/site-packages/SQLalchemy-1.1.4-py3.5-linux-x86_64.egg/sqlalchemy
08/15/2017 11:05:26 INFO: Gathering files to ingest
08/15/2017 11:05:26 INFO: Gathering files to ingest
08/15/2017 11:06:01 INFO: J&zhlzvu: Updated HRC_fitt_1 & table.
08/15/2017 11:06:01 INFO: J&zhlzvu: Updated HRC_fitt_3 table.
08/15/2017 11:06:01 INFO: J&zhlz
```

Fig. 18: An example log file for the ingestion of a single file (j8zh21xv).

LICENSE: A BSD 3-Clause license, which states that the acsql package is an open source software package and may be used and redistributed.

README.md: A README file that describes how to install and use the acsql package.

MANIFEST.in: A list of static files to be included in the tarball file when the user installs the acsql package.

setup.py: The acsql package installtion script. Executing this script with python setup.py install installs the package into the software environment.

paper/: A subdirectory which contains all materials used for the creation of this paper.

presentation/: A subdirectory which contains all materials used for the creation of the COSC 880 presentation.

docs/: A subdirectory which contains all materials used for the creation of the sphinx API docuemntation hosted on Read the Docs (see Section 3.2).

Makefile: A make script that is used to build the sphinx API documentation from the source reStructured Text files (see below) (see Section 3.2).

requirements.rst: A list of acsql package dependencies, used by Read the Docs to build a virtual machine that constructs the resulting html doc pages (see Section 3.2).

source/: A subdirectory that contains all of the reStructured Text files used for building the sphinx API docuemntation, one .rst file per subpackage, including a master index.rst file (see Section 3.2).

acsql/: A subdirectory that contains all Python code that is part of the official acsql Library. This is the top level of

the importable acsql package.

__init__.py: A Python file that indicates that the subdirectory is part of the overall acsql package.

database/: The database subpackage, containing Python modules that pertain to the acsql database (see Sections 3.5 and 3.6).

database_interface.py: The Python module for constructing and connecting to the acsql database (see Section 3.6).

make_tabledefs.py: The Python module for creating the table definition text files (see Section 3.6)

queries.py: A Python module that contains several examples of queries that can be used with the acsql database.

 $\begin{tabular}{ll} reset_database.py: A Python module that allows the user to 'reset' the acsql database (i.e. drop all tables, then create all tables). \\ \end{tabular}$

table_definitions: A subdirectory containing all of the <detector>_<filetype>_<extension> text files, each of which contain a list of header keys along with their datatypes (see Section 3.6).

update_tabledefs.py: A Python module that allows the user to update the table_definitions text files with new header keywords (see Section 3.6).

ingest/: The ingest subpackage, containing Python modules for ingesting new data into the acsql system, including database updates and the creation of JPEGs/Thumbnails (see Section 3.7).

ingest.py: A Python module for performing the ingestion of a single file (see Section 4.7).

make_file_dict.py: A Python module for creating a file_dict for an individual file (see Section 3.7).

make_jpeg.py: A Python module for creating a JPEG image from an individal file (see Section 3.7).

make_thumbnail.py: A Python module for creating a Thumbnail image from an individal JPEG (see Section 3.7).

scripts/: The scripts subpackage, containing Python modules for ingesting multiple files from the acsql filesystem, as well as storage place for possible future instrument calibration and monitoring routines.

ingest_production: The Python module for ingesting new ACS data as it becomes publicly available, intended to be executed periodically.

utils/: The utils subpackage, containing Python modules that are useful for general acsql operations (e.g. configuring loggging, supplying hard-coded instrument configurations, etc.) as well as a configuration file for storing sensitive credentials and directory locations.

config.yaml: A text file containing hard-coded user-specific directory locations and acsql database credentials. Specifially, it contains values for the acsql database connection_string, as well as locations for the filesystem, log_dir, jpeg_dir, and thumbnail_dir. The contents of the config.yaml file can be imported via the utils.utils.SETTINGS dictionary.

utils.py: A Python module containing various functions that are generally useful for acsql operations, such as configuring logging, determining if a database entry requires an insert or an update, and hard-coded Python variables that reflect instrument/system configurations.

website/: The website subpackage, containing Python modules that are used in the construction and operations of the acsql web application (see Section 3.9).

acsql_webapp.py: The main Python module for running the acsql web application, using the Python Flask web framework (see Section 3.9).

data_containers.py: The Python module that contains various functions for reterning various data to be used by the acsql web application (see Section 3.9).

form_options.py: A Python module that stores form data for the database query portion of the acsql web application (see Section 3.9)

query_form.py: A Python module that contains class objects for building the query form for the database query portion of the acsql web application (see Section 3.9)

query_lib.py: A Python module that contains various functions to support the querying of the acsql databse

through the acsql web application.

static/: A subdirectory containing static materials used by the acsql web application, such as CSS templates (i.e. css/, JavaScript functions (i.e. js/, and symbolic links to the JPEGs and Thumbnails hosted on the web application (i.e. img/.

templates/: A subdirectory containing HTML templates used to render the various webpages of the acsql web application, one for each page.

For further details on each Python module within the acsql package, readers are encouraged to view the official API documentation hosted at http://acsql.readthedocs.io/.

4 RESULTS

Topics to discuss: 1. GitHub repository 2. ReadTheDocs documentation repository 3. Quantification of Database records 4. Quantification of Code repository 5. Website location

5 CONCLUSION

The conclusion goes here.

6 Discussion

Topics to discuss:

1. Possible simplification based on MAST archive 2. Possible extensions to other insturments

APPENDIX A

ACSQL CODE

Appendix one text goes here.

ACKNOWLEDGMENTS

The authors would like to thank...

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