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

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Abstract—What is an abstract, really?

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 for any publicly-available dataset via ftp, sftp, or DVD by mail². The ACS data, like most all other astronomical data, are stored in the Flexible Image Transport System (FITS) filetype (FITS, 2008). This filetype has several unique characteristics, as will be discuessed in section 1.1.

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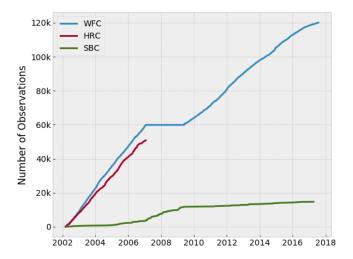


Fig. 1. The number of observations over time for each of the three detectors on ACS.

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 "Quicklook" 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

^{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

^{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.

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 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

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

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 multi-dimensional 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).

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.

```
bits per data value
                                                 0 / number of data axes
T / File may contain standard extensions
6 / Number of standard extensions
                                                    / number of scandard extensions
/ image is in group format
/ date this file was written (yyyy-mm-dd)
/ type of data found in data file
ILENAME= 'j6mf16lhq_raw.fits
ILETYPE= 'SCI '
TELESCOP= 'HST'
INSTRUME= 'ACS
EQUINOX =
                                        / 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
                       / 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
    INV_L= '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)
ATE-0BS= '2003-01-27'
```

Fig. 2. An example header.

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 <code>j6mf16lhq</code> 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.).

 $TIME ext{-}OBS$ - 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.

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.

```
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 <code>get_proposal_type</code> function from <code>acsql.ingest.ingest</code>.

```
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 , 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. 4. The ${\tt readthedocs}$ documentation for the ${\tt acsql}$ example function seen in Figure N.

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-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.

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

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 ~40 TB of storage space. Note that the file sizes across the detectors and across the various filetypes 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 RAW, FLT, and FLC 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 twodimensional 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

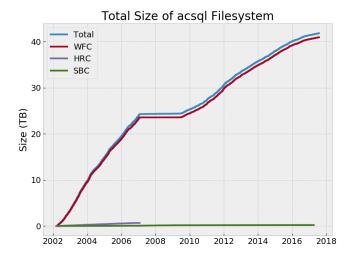


Fig. 6. The size of the ${\tt acsql}$ filesystem as a function of observation date.

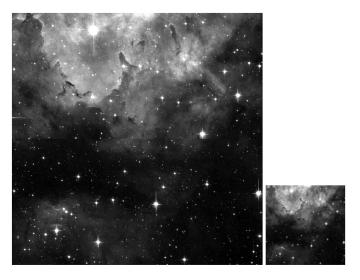


Fig. 7. An example of a JPEG image (left) and its corresponding thumbnail image using example dataset jcs718koq.

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.

```
jpegs/
    12780/
    jbw901jtq_flc.jpg
    jbw901jtq_raw.jpg
    jbw901jxq_flc.jpg
    jbw901jxq_flc.jpg
    jbw901jxq_raw.jpg
    jbw901jxq_raw.jpg
    jbw901jxq_raw.jpg
    jbw901jxq_raw.jpg
    jbx101f5q_flc.jpg
    jbx101f5q_flc.jpg
    jbx101f5q_raw.jpg
    jbx101f7q_flc.jpg
    jbx101f7q_flc.jpg
    jbx101f7q_raw.jpg
    12783/
    ...
    12787/
    ...
    ...

thumbnails/
    jbw901jtq_flt.thumb
    jbw901jtq_flt.thumb
    jbw901jtq_flt.thumb
    jbw101f5q_flt.thumb
    jbx101f5q_flt.thumb
    ...
    12783/
    ...
    ...
```

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

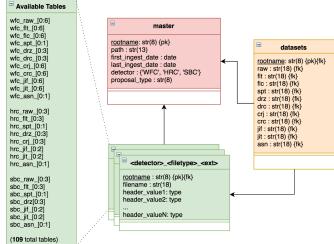


Fig. 9. The relational database schema for the acsql database.

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. Its primary key is the first 8 characters of the 9-character rootname for the particular observation (recall from sec-

tion 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 key functions and classes were acsql that used to construct database (all of which can be found 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 the sqlalchemy.create_engine method, which requires a user-supplied connection_string. The connection_string is a string that contains information

```
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
       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 <code>load_conenction</code> function, which is used to build a connection to the <code>acsql</code> database

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 discussed in section 3.9).

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).

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,

Fig. 11. The class definition for constructing the ${\tt master}$ table via ${\tt SQLAlchemy}.$

```
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 ${\tt datasets}$ table via ${\tt SQLAlchemy}.$

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). 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 for each header table, based on the given class_name

```
def orm_factory(class_name):
   """Create a SQLAlchemy ORM Classes with the given ``class_name``.
   Parameters
   class name : str
       The name of the class to be created
   class : obj
       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 ${\tt orm_factory}$ function, used to define class definitions for header tables.

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 (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 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, cre-

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

```
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.
   Returns
   data dict : dict
       A dictionary containing the ORM definitions, now with header
       definitions added.
   special_keywords = ['RULEFILE', 'FWERROR', 'FW2ERROR', 'PROPTTL1',
                       'TARDESCR', 'QUALCOM2']
   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)
   return data dict
```

Fig. 14. The ${\tt define_columns}$ function, used to define columns used in the header tables.

ating the database tables on the MySQL server is accomplished by executing the base.metadata.create_all() method.

- 3.7 Data ingestion software
- 3.8 Website:
- 3.9 acsql Library

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

```
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.

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

Appendix one text goes here.

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