



Astrolabe: Curating, Linking, and Computing Astronomy's Dark Data

P. Bryan Heidorn¹ , Gretchen R. Stahlman¹ , and Julie Steffen²

¹ University of Arizona School of Information, Harvill Building 4th Floor, 1103 East 2nd Street, Tucson, AZ 85721, USA

heidorn@email.arizona.edu, gstahlman@email.arizona.edu

² American Astronomical Society, Washington, DC, USA; julie.steffen@aas.org

Received 2017 October 19; revised 2018 February 25; accepted 2018 February 26; published 2018 May 11

Abstract

Where appropriate repositories are not available to support all relevant astronomical data products, data can fall into darkness: unseen and unavailable for future reference and reuse. Some data in this category are legacy or old data, but newer data sets are also often uncurated and could remain dark. This paper provides a description of the design motivation and development of Astrolabe, a cyberinfrastructure project that addresses a set of community recommendations for locating and ensuring the long-term curation of dark or otherwise at-risk data and integrated computing. This paper also describes the outcomes of the series of community workshops that informed creation of Astrolabe. According to participants in these workshops, much astronomical dark data currently exist that are not curated elsewhere, as well as software that can only be executed by a few individuals and therefore becomes unusable because of changes in computing platforms. Astronomical research questions and challenges would be better addressed with integrated data and computational resources that fall outside the scope of existing observatory and space mission projects. As a solution, the design of the Astrolabe system is aimed at developing new resources for management of astronomical data. The project is based in CyVerse cyberinfrastructure technology and is a collaboration between the University of Arizona and the American Astronomical Society. Overall, the project aims to support open access to research data by leveraging existing cyberinfrastructure resources and promoting scientific discovery by making potentially useful data available to the astronomical community, in a computable format.

Key words: astronomical databases

1. Introduction

Research in astronomy has changed dramatically over the past century: while astronomers 100 years ago would be intimately familiar with a telescope itself, spending many hours observing the sky and making observations in direct contact and collaboration with the instrument, modern astronomers rely heavily on data output, and often on combinations of telescope and archival data (McCray 2004, 2014). To successfully answer research questions across astronomical subfields, both old and new data are useful. However, research programs outside of major observatories often lack the infrastructure for proper data management systems to keep older data relevant, and the NSF-funded report *Future Directions for NSF Advanced Computing Infrastructure to Support U.S. Science and Engineering in 2017–2020* points out that an NSF-wide cyberinfrastructure strategy or program does not exist to support disciplinary or cross-disciplinary data sharing and preservation (National Academies 2016). Where appropriate repositories are not available to support all relevant astronomical data products, data can fall into darkness: unseen and unavailable for future reference and reuse. Some data in this category are legacy or old data, but newer data sets are also often uncurated (Hanisch et al. 2017) and could remain “dark” (Heidorn 2008).

This paper provides a description of Astrolabe, a cyberinfrastructure project that addresses a set of community recommendations for locating and ensuring the long-term curation of otherwise dark or at-risk data and integrated computing, in coordination with Open Science initiatives (Open Science Collaboration 2012). According to participants in a series of workshops including astronomers, cyberinfrastructure specialists, computational scientists, and administrators

(see Section 5), many astronomical dark data currently exist that are not curated. NASA astronomical missions and large Earth-based instruments include planning for data management, and this efficient project-management approach to instrumentation serves the astronomical community well. However, scientists routinely find new ways to analyze, merge, and manipulate data, creating new data sets that do not meet the collection requirements of the original projects. This leads to weaknesses in terms of discoverability, compatibility, and storage of derivative data. Workshop participants further reported that some software can only be executed by a few individuals, and therefore becomes unusable because of changes in computing platforms. Astronomical research questions and challenges would be better addressed with integrated data and computational resources that span facilities and mission archives.

As a solution, the design of the Astrolabe system is based on input from this group of experts, aimed at developing new resources for management of previously uncurated astronomical data, with an emphasis on small, old, and sometimes heterogeneous data sets. Workshop participants recognized that some newer and larger data sets are also uncurated or lack long-term data management planning. As described in the sections below, workshop participants provide insight into the data and cyberinfrastructure needs of the astronomical community, including data collection, analysis, and publication practices. Recommendations include building on existing astronomy data systems, as well as borrowing heavily from data and computation systems built for other sciences, with an objective of enabling and accelerating new science through improved curation of both older and newer data. In response to this feedback, we decided to build

Astrolabe on the CyVerse cyberinfrastructure,³ which was already put into place by the biology community, as detailed later in this paper.

Astrolabe collaborates with the American Astronomical Society (AAS); through this partnership, participants in the Astrolabe workshops suggested that Astrolabe should host data corresponding to published research not currently curated by trusted repositories. The Astrolabe team is also seeking high-value data sets (as judged by the astronomical community) that are not associated with publications. Overall, the project aims to support open access to research data by leveraging existing cyberinfrastructure resources, and to promote scientific discovery by making potentially useful data broadly available to the astronomical community, in a computable format. The purpose of this paper is twofold: (1) to describe the data management requirements, identified by expert participants in a series of workshops, necessary for successful curation of dark data in astronomy; and (2) to describe the development of Astrolabe in response to these identified requirements. The paper is outlined as follows. Background information on the dark data problem and open science initiatives is provided in Sections 2 and 3, followed by a description of the Astrolabe system and development activities in Section 4. Sections 5 and 6 give a presentation of the workshops and their outcomes, with a related discussion in Section 7. Finally, Section 8 concludes the paper with a summary and directions for future work.

2. Dark Data in the Long Tail of Astronomy

Keeping track of dark data in astronomy is important for future scientific discovery. As an illustrative analogy from economics, consider the emergence of Netflix online while Blockbuster Video store fronts closed around the United States. Long Tail economics theory (Anderson 2004) notes that markets for goods are frequently based on a small number of high-volume items. In the age of physical brick-and-mortar stores, the goal is to keep high-volume items on the shelves and available for purchase, as low-volume items take up space and reduce sales. Best-selling popular movies stocked on Blockbuster shelves are examples of high-volume items, while the many more less-popular films and documentaries are low-volume items, warehoused by Netflix Online. Netflix was able to economically provide access to low-volume items through the Internet, without incurring the costs of physical shelf space in a brick-and-mortar store. Anderson represents this Long Tail in a graph much like the one depicted in Figure 1 below. The total area under the right part of the curve—the low-volume items—represent significant economic opportunity.

Long Tail thinking in science was proposed by (Heidorn 2008), who performed an exploration of NSF-funded research grants across disciplines, showing that the largest 20% of projects funded in 2007 received more than 50% of the total funds awarded in that year. While the 2008 paper analyzed all of the NSF programs in 2007, Figure 1 shows the distribution of funds in NSF astronomy programs in 2016.⁴ The astronomy funding distribution is very similar to that of all programs in NSF in 2007, with a few very large projects and many smaller projects. Heidorn (2008) noted that this top 20% has

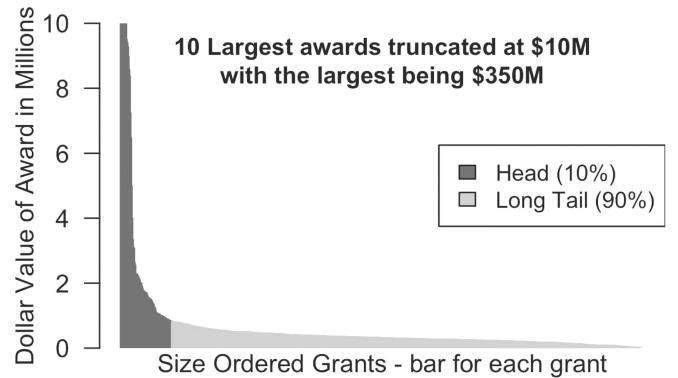


Figure 1. Long Tail of NSF astronomy research funding. The data shown here represents 855 astronomy and astrophysics projects. The top 10% of projects received between 350,000,000 to about 830,000 dollars. The bottom 90% of projects received between about 830,000,000 to 5,000 dollars. The highest-funded projects require a sophisticated data management plan, meaning the Long Tail of research funding likely translates to a Long Tail of dark (uncurated) data.

well-curated data through access to curation resources and budgeting for data management in engineering and project plans, and that the uncurated data distributed throughout the Long Tail—i.e., the lower annual funding part of this distribution—represent potentially valuable information that could be relevant to current research projects; curation of all data, including those from smaller projects, may lead to new discoveries. While there are resources to provide long-term homes for some of these data, much of the data from small projects end up without long-term curation. This view is supported by reports of scientists from the workshops discussed in later sections.

If data are not stored and organized efficiently, scientists will not be able to find and use the data, rendering those data essentially “dark.” According to Heidorn (2008), dark data requires unique curation strategies to capture and manage data that are not easily accessed by potential users. Dark data in the Long Tail are typically: heterogeneous; hand-generated; created through unique procedures; curated by individuals; often archived in personal or institutional repositories; not maintained; obscured or protected; seldom reused; and currently unnoticed. This Long Tail data is more likely to be lost over time, in a process referred to in the field of ecology as “data decay” (Michener et al. 1997). Without initial good curation, the legacy data in the Long Tail is more likely to be lost than the legacy data from larger and originally well-curated projects. So, small project size and the associated lack of resources for metadata and curation, as well as time (legacy), can all combine to threaten data availability. The participants in the workshops presented here largely agreed with this characterization.

Heidorn (2008) states that it is necessary to understand dark data in order to better manage it, a notion that led to and guided the two workshops held in 2015 and 2016 (Stahlman et al. 2018). In the discipline of astronomy, the Long Tail is distinct and multifaceted. Borgman (2015) explains that data volume in astronomy is now growing at a particularly tremendous rate with each new instrument. However, many astronomers implement small research projects, often with specialized instruments for data collection, or rely on derivative and/or theoretical analyses using data available from mission or observatory archives. These data are often not made publicly

³ <http://www.cyverse.org>

⁴ A log transform of this graph would yield a nearly straight line, and thus obscure the tail.

available, frequently due to the complexity of the data sets themselves, lack of data management skills, and because the utility of data for other researchers may not outweigh the difficulty in providing metadata and stable storage and access (Edwards et al. 2011; Borgman 2015). While astronomy is increasingly a “big data” field, “little” or “small” science—which is the primary focus of the Long Tail theory—is typically hypothesis-driven and led by only one principal investigator rather than a large team (Cragin et al. 2010). Developments in astronomical research toward team-based collaborations and rapid generation of data in astronomy do not necessarily preclude the ongoing presence of “small science” in astronomy, where scientists may generate and hold innovative data in private collections.

3. Towards Open Data and Open Science

Major funding agencies increasingly recognize the importance that public access to research output has in facilitating knowledge production, particularly when this research is funded through public support. Many funding agencies now require proposers to provide plans for data management, and also to upload copies of resulting journal articles to public archives. The National Science Foundation (NSF 2015), a leading source of funding for astronomical research and instrument construction in United States, recently published a vision to explore how to improve public access to data, including storage, preservation, discoverability, and reuse, with a focus on data and publications associated with federally funded scientific research. In response to this plan, NSF sponsored a series of workshops within the Directorate for Mathematical and Physical Sciences to obtain feedback from the research community, and to produce recommendations for NSF on realizing this vision. Through this process, the research community is exploring existing initiatives exemplifying best practices and models that could be adapted by NSF (Hanisch et al. 2017), indicating that the topic of open data and supporting cyberinfrastructure is of cutting-edge importance now and for the future of astronomy.

Repositories exist for raw and calibrated astronomy data associated with large facilities, which regularly provide Level 1, 2, and 3 data products. While there is some variation of definitions, Herschel Data Products⁵ defines Level 1 as detector readouts calibrated and converted to physical units. Level 2 products are processed for scientific applications and often reviewed by humans. Level 3 are produced from merged level 2 products. Curated repositories include, for example, the National Optical Astronomy Observatory’s DataLab,⁶ which connects catalog objects with NOAO images and additional services for data analysis. Other repositories, such as the NASA/IPAC Extragalactic Database (NED)⁷ and VizieR,⁸ are trusted collections of catalogs, tables, and images. Along with institutional repositories, Harvard’s Dataverse⁹ repository and CERN’s Zenodo¹⁰ are resources, currently used by some astronomers, offering intermediate data products and data associated with publications that are

not archived elsewhere. The value of data archives has been demonstrated, for example, by White et al. (2009), where the number of papers based on archival data from the *Hubble Space Telescope* exceeded the number of non-archival publications. While many data are preserved and accessible in well-curated repositories, as described above, many other data products—including data derived from analyses of mission data, synthesis of data from multiple missions or instruments, and supporting published research—often remain hidden and require development of resources for data access and preservation throughout the lifecycles of these data (Conrad et al. 2017).

The National Virtual Observatory (NVO) initiative was conceptualized in a 2001 white paper (Brunner et al. 2001) and evolved into a series of Virtual Observatory (VO) projects and products. The NVO was envisioned as a semantic web of ontologically linked knowledge encompassing the full research lifecycle of archived raw and derived astronomical data, computation, and software, as well as project proposals and publications (Brunner et al. 2001; Hanisch et al. 2007; Accomazzi 2011). Furthermore, the NVO was expected to collaborate with academic research libraries for long-term curation (Choudhury et al. 2008). VO technology and standards are now actively used throughout the international astronomical community, representing an opportunity for linking data in the Long Tail to corresponding literature and broadly facilitating discoverability. Integral to supporting astronomical research, the Astrophysics Data System (ADS) is a robust open-access tool, indexing nearly all astronomical publications and associated metadata, including links to data when available (Accomazzi & Dave 2011; Accomazzi et al. 2016), and with capabilities for visualizing citation and collaboration patterns (Henneken et al. 2009). Using the ADS, Henneken & Accomazzi (2011) show that publications with links to data are more highly cited than publications that do not link to data, an important finding for open data and data-sharing initiatives (Henneken 2015).

4. Astrolabe

As shown in previous sections, where appropriate repositories do not exist, data can fall into darkness. A functional repository must: make data easy to ingest, format, and normalize; index it for discoverability; and transform, visualize, and deliver data for reuse. Fortunately, not all of this functionality must be built from scratch. Astrolabe is a new system for storage and analysis of previously uncurated astronomical data, built using the existing CyVerse¹¹ cyberinfrastructure platform. Many of the Astrolabe design elements within CyVerse have been created in direct response to recommendations of scientists in the workshops presented below. As a community-oriented and multi-disciplinary cyberinfrastructure designed to support open science, CyVerse collaborates with other repositories to ensure interoperability and ongoing development and deployment of its versatile technology across institutions (Lenhardt et al. 2016). Flexible cyberinfrastructure (i.e., infrastructure based on distributed computer, information, and communication technology) is key to these objectives. CyVerse provides a cloud computing environment that helps to free end users from the necessity to install and configure sometimes-complex systems of programs on their laboratory computers. Originally known as the iPlant

⁵ <https://www.cosmos.esa.int/web/herschel/data-products-overview>

⁶ <http://datalab.noao.edu/>

⁷ <https://ned.ipac.caltech.edu/>

⁸ <http://vizier.u-strasbg.fr/viz-bin/VizieR>

⁹ <https://dataverse.harvard.edu/>

¹⁰ <https://zenodo.org/>

¹¹ cyverse.org

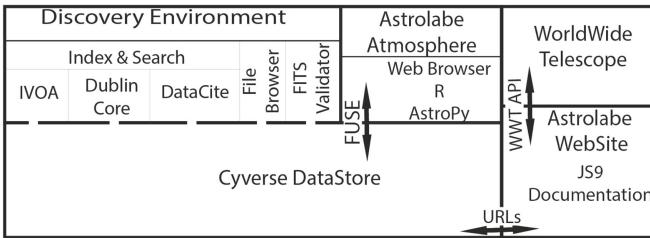


Figure 2. Astrolabe system diagram.

Collaborative and supported by NSF since 2008 (Atkins 2003), CyVerse provides computation, data storage, portability, and data federation (controlled copies) through iRODS¹² data management software. Links to supercomputing resources for cloud-based data analysis and software tool development provide scalability. These features have attracted more than 31,000 active users and nearly 3,000 terabytes of user data (CyVerse 2017). Initially designed for the biological community, CyVerse is expanding its services to other disciplines. Projects from a variety of disciplines across natural and life science domains are now based in CyVerse technology. This type of computation supports software and data reusability, repeatability of computations, and overall validity in science.

Astronomy has produced a rich set of processing resources and standards. However, these tools and standards sometimes require very particular processing environments, making it difficult to develop an integrated operating environment for astronomical data. Astrolabe workshop participants suggested using CyVerse as one of the platforms for delivering astronomical repository and processing services, particularly because the CyVerse cyberinfrastructure elements adapt well to the needs of a variety of scientific disciplines, including astronomy. CyVerse supports Docker¹³ technology, which allows developers to package together all of the languages, libraries, and other tools, along with their dependencies, and wrap them into portable containers, thus isolating users from the complexity of the relationships. CyVerse also supports iRODS to allow for easy control of workflows, even for large data sets, as well as movement of data between repositories, supporting long-term sustainability. Furthermore, CyVerse provides access to scalable computing with relatively easy access to high-performance computing for end users.

The Virtual Observatory project provides useful guidelines for Astrolabe development, including basic discovery functionality and data federation to allow search across repositories. Generally, raw data generated by an instrument must be processed into a form that is more directly useful for astronomers, and this involves processes such as calibration and registration, as well as maintenance of metadata. There are multiple data formats in astronomy, but a few formats dominate (Greenfield et al. 2015), including Flexible Image Transport System (FITS) (Grosbol 1988) and Hierarchical Data Format 5 (HDF 5) (Folk et al. 2011), so Astrolabe is creating workflows to handle processing of these data types. We are also currently directing Astrolabe efforts toward linking tools such as AstroPy, PyRAF, CarPy, JS9, and Montage to CyVerse to provide astronomers with easy access to these applications,

within the repository environment and through an attractive user portal.

Below, we discuss the Astrolabe/CyVerse ecosystem (depicted in Figure 2) and briefly outline how an astronomer can begin using Astrolabe.

4.1. The Astrolabe-CyVerse Eco-system

Astronomers can use Astrolabe to store and share data, and importantly, they can process data in a cloud environment. The Astrolabe repository uses CyVerse as a storage and analysis environment for heterogeneous astronomy data, and provides digital object identifiers (DOIs) to link data to corresponding literature and to other established databases for overall transparency and reuse of research outcomes in Physical Sciences. Furthermore, Astrolabe is developing WorldWide Telescope (WWT) open-source software as a scalable repository front end and tool for visualizing spatial data, such as astronomical or planetary images, on the sky and in three dimensions; see Rosenfield (2018). Astronomers may interact with Astrolabe through the following established links to the Cyverse Data Store (which holds the actionable data for Astrolabe). (1) Users may enter the Astrolabe website and directly access select items in the Astrolabe repository or use our web services in a standalone manner (e.g., a JS9 implementation that connects to WorldWide Telescope). (2) Users may enter through the CyVerse Discovery Environment (DE), to upload data to the Astrolabe repository (using the CyVerse DataStore), search metadata to find items in the repository, and/or run Dockerized applications in the DE. Users can also analyze data in the repository using astronomical software with cloud services (Atmosphere). We discuss each of these products and services below.

CyVerse DataStore—The CyVerse DataStore is a cloud-based storage system with parallel IO for rapid movement of data. Astrolabe users have fine-grained control of access through authentication services. Astrolabe has multiple terabytes of storage, and each individual Astrolabe user is allocated an additional 100 gigabytes (current transfer rates are about 1 gigabyte per 18 s). The DataStore can be accessed directly using industry standard protocols such as iRODS (iCommands),¹⁴ and via batch mode using file transfer tools such as CyberDuck,¹⁵ which supports iRODS. Astrolabe data exists in two main partitions—as a private space, and as public “Community Data” that will be discussed in the following description of the Discovery Environment.

Discovery Environment—The Discovery Environment (DE) is a web interface that provides access to data and applications that can run on the CyVerse hardware cluster or on Texas Advanced Computing Center (TACC) supercomputers. Data can be accessed via browsing or through searching over multiple metadata frameworks with CyVerse ElasticSearch.¹⁶ Astrolabe currently uses three metadata indexing and search templates: Dublin Core¹⁷ and DataCite¹⁸ (both are native to CyVerse), as well as a new metadata template created for

¹⁴ <https://irods.org>

¹⁵ <https://cyberduck.io>

¹⁶ <https://pods.implantcollaborative.org/wiki/display/-bjoyce3/ElasticSearch+Integration+in+CyVerse>

¹⁷ <http://dublincore.org/>

¹⁸ <https://www.datacite.org/>

Astrolabe following the IVOA ObsCore¹⁹ standard and the FITS and WCS standards.²⁰ The Astrolabe metadata template can be populated directly from imported FITS files by using a Discovery Environment FITS Metadata Extractor application. The Astrolabe metadata template and the DataCite template each include the capability to select Unified Astronomy Thesaurus (Frey & Accomazzi 2018) terms as keywords in the subject field. DOIs can be requested after assigning DataCite metadata to data sets and/or files. To ensure the integrity of this process, Astrolabe is also participating in ongoing discussions with other institutions and projects about best practices for assigning DOIs to astronomical data; see Novacec et al. (2018). DataCite metadata records and DOIs are only assigned to collections of data that are considered to be permanent and unchanging, with appropriately detailed descriptions. Scientists can also create and share applications (apps) in the DE (the FITS Metadata Extractor application is an example). Hundreds of apps currently exist, though most are designed for biology. All DE apps are “containerized” with Docker, allowing them to execute on multiple platforms.

Atmosphere Virtual Machine—An Astrolabe Atmosphere virtual machine has been pre-configured with astronomical data analysis software through CyVerse’s “Atmosphere” environment. An Astrolabe user can run an instance of this machine, perform tasks such as editing images with JS9, and then conveniently save the outputs back into the user’s Astrolabe CyVerse DataStore account. Astrolabe Atmosphere users can boot servers from 1 to 16 processors, memory from 4 to 128 gigabytes of RAM, and 30–1200 gigabytes of disk depending on the processing needs. The Atmosphere environment is being deployed in the NSF-funded JetStream project (Towns et al. 2014; Stewart et al. 2015). Planning for JetStream compatibility allows for sustainability and potential for migration to an environment that can provide hundreds of processors. Astrolabe’s Atmosphere implementations of Astropy and R-Suite are under development. Other astronomical software, such as PyRAF/IRAF, will be added based on user demand. Astrolabe’s Atmosphere runs a web browser that can execute the modified JS9 implementation running on the Astrolabe web server discussed below and could run other web services as well. This allows the DataStore to be accessed directly through JS9. A detailed workflow for using this software is provided below.

Astrolabe Web Server—The Astrolabe web server is hosted on the CyVerse Cloud. The website²¹ hosts information on the project and hyperlinks to select collections in the DataStore, as well as an instance of the JS9 public API implementation.²² JS9 can be used to edit FITS metadata and images for visualization in WWT or direct download. JS9 can also create FITS files from other image types, such as a .jpg or .png. Furthermore, this JS9 implementation can run using a web browser in the Astrolabe Atmosphere environment to allow additional processing and direct access to files in the CyVerse DataStore.

Astrolabe does not use all CyVerse functionality, and other services may be added to Astrolabe. The process of accessing and utilizing Astrolabe through CyVerse is discussed in the next section.

¹⁹ <http://ivoa.net/documents/ObsCore/index.html>

²⁰ <https://fits.gsfc.nasa.gov/>

²¹ <http://astrolabe.arizona.edu>

²² <https://js9.si.edu/js9/help/publicapi.html>

4.2. Using Astrolabe

As mentioned above, CyVerse was initially developed to support life sciences, but has since endeavored to expand to other disciplines in the United States research community. Prior to the inception of Astrolabe, CyVerse was used by only a few individual astronomers for particular instruments and data analysis tasks. Astrolabe now endeavors to develop attractive and useful tools, and to widely connect astronomers to CyVerse resources, augmenting an already-substantial multidisciplinary user community with astronomical researchers and data. As a first step toward creating a fully integrated web portal to CyVerse, current and potential users can learn about Astrolabe features and how to access CyVerse through the project’s website: <http://astrolabe.arizona.edu>.

Detailed instructions on using Astrolabe are provided on the Astrolabe website. To summarize, users must first create a CyVerse account at <http://www.cyverse.org/> to join the Astrolabe community. The users can then login to the Discovery Environment, create a subdirectory, upload data, set permissions to share data (or keep some private), add metadata for discoverability, edit images or FITS headers, and view images in the WorldWide Telescope.

Pilot Data—Astrolabe is actively seeking users and data sets. At the time of this publication, the project has successfully ingested several data sets, including the AAS CD-ROM Series, which was launched in 1993 and includes nine volumes of data corresponding to publications between 1993 and 1997. These data are primarily catalogs and tables, and more than 300 folders have been indexed in CyVerse with preliminary DataCite (DOI) metadata, searchable with CyVerse’s beta search function.

5. Astrolabe Community Workshops

Informing the creation, design, and implementation of Astrolabe, two separate workshops were held for communities of researchers, computational staff, and administrators in astronomy. Participants in both workshops were primarily recruited from the University of Arizona, National Optical Astronomy Observatory (NOAO), the AAS, and CyVerse. Data collected from the workshops include: field notes taken by workshop organizers using a collaborative note-taking platform; preliminary surveys administered online; audio recordings of presentations, group discussions and breakout sessions; and transcription of key dialogue. This material was analyzed qualitatively by coding for themes and sub-themes. The workshop methods and results of these analyses are summarized below.

5.1. Preliminary Surveys

Guiding the design of the two Astrolabe workshops and session topics, a short, informal, open-ended online survey was sent to all workshop invitees prior to each event. Questions targeted: technical and social issues associated with uncurated astronomy data; strengths and weaknesses of existing data repositories; science use cases for a new astronomy repository; opportunities for education and outreach; and opportunities and challenges associated with the proposed project. Respondents provided substantial text-based feedback and inquiries; key themes were identified prior to each workshop and addressed through workshop breakout sessions.

5.2. Astrolabe Workshop Design

Invitees to the 2015 “Arizona Astronomical Data Hub—AADH” (the initial tentative name for the Astrolabe project) workshop included astronomy and data science experts, primarily affiliated with University of Arizona, NOAO, and the AAS, with several invited speakers, and remote participation by individuals at the Harvard-Smithsonian Center for Astrophysics. The workshop itself loosely followed the Delphi Method of achieving group consensus (Keeney et al. 2001). Thirty-two participants (remote and in-person) and facilitators were organized into groups based on expertise, ensuring that each group contained a representative mix of astronomers, data scientists, librarians, and educators/administrators. Following an initial introduction by guest speakers, four breakout sessions were held over two days, each immediately followed by whole-group report-out sessions. Detailed notes were taken by each facilitator and by participants using a collaborative note-taking platform, and invited talks and whole-group discussions were recorded; this material was analyzed and coded to identify the following broad themes and detailed objectives for the Astrolabe project.

A follow-on 2016 Astrolabe workshop included twenty-two local and nonlocal participants. The group was comprised of research astronomers, University of Arizona and CyVerse representatives, and others with expertise related to astronomy data and publishing, along with five workshop organizers. Just prior to the Astrolabe workshop, the first meeting of the Astrolabe Advisory Board was convened. Board members broadly discussed the research challenges influencing the creation of Astrolabe, as well as the general function and operation of the board. Along with the preliminary online survey that was distributed to participants, this meeting helped to inform the design of the workshop’s three breakout sessions and related group discussions.

The 2016 workshop began with a welcome and introduction from the AAS. Workshop participants were then updated on Astrolabe, the project’s activities and progress over the previous year, and the workshop structure and objectives. A series of invited talks provided an overview of: the resources available through CyVerse; the UAT; a general discussion of data management and visualization, as well as publishing in astronomy; and a demonstration of the functionality and new searching capabilities of the ADS index of astronomy literature. The first day’s afternoon session asked four breakout groups to discuss the most important and practical solutions, data types, and astronomy subfields to direct focus for incorporating dark data that is both useful and accessible into Astrolabe. When the workshop reconvened the following morning, each breakout group briefly discussed the prior day’s conclusions and then reported its findings. The key areas of focus identified by the four breakout groups were listed, and participants voted on the four most critical and practical ones. Subsequent breakout sessions were organized around four topic areas recommended for the immediate attention of Astrolabe. Workshop participants explored these topics in depth during a breakout session directed toward the functional requirements that would be necessary to address each topic from both a user and developer perspective. The final session was condensed to three breakout groups to discuss funding opportunities and overall strategies for the sustainable growth of Astrolabe. Several additional invited talks on the second afternoon provided further context

for Astrolabe, through a discussion of the strengths and weaknesses of several case studies. The workshop concluded with a final whole-group discussion.

The key topics identified below were sometimes discussed in more than one of the two surveys or workshops, but they are presented only once for simplicity. The importance of each key topic is supported by relevant participant feedback. This validation of the dark data problem (and surrounding issues and opportunities) is evident in recorded responses to the pre-workshop survey sent to all participants (8/32 responses in 2015, 7/22 responses in 2016). Survey responses are interspersed through the following section, to illustrate the qualitative insights derived from the 2015 and 2016 workshops.

6. Workshop Outcomes

6.1. 2015 Astrolabe Workshop Outcomes

The following broad themes and detailed objectives were identified for the Astrolabe project, with confirmation from pre-workshop survey responses:

6.1.1. Identify Mission and Clear Science Use Cases

One or more detailed science use cases are essential to obtaining support and creating a viable new repository for dark or uncurated data in astronomy. Therefore, establishing and documenting potential science cases are important objectives of this project. Workshop discussions indicated that certain existing survey data sets are in danger of becoming dark data and could illustrate the utility of a new astronomy data repository. For example:

Each snapshot of the universe is unique. Data may represent the only image or spectrum corresponding to a particular timestamp in a particular direction, certainly from a particular observatory location. Simply collecting and protecting the data should be the most fundamental goal. (Anonymous Respondent, 2016 pre-workshop survey).

In addition, it was recommended that the project mine the published literature for references to uncurated data sets. Key among the suggested use cases is the time domain, where new discoveries lead to interest in old observations, which could be dark data:

The generic argument for curating every direct image ever taken is the potential value for time domain studies—searching backwards in time for historical outbursts of unstable systems, long-term stellar variability for exoplanet hosts, multiple supernovae in the same galaxy, or coming soon—optical counterparts to gravitational wave signals and/or gamma-ray bursts extracted from the multi-messenger data stream after the fact. (Anonymous Respondent, 2016 pre-workshop survey).

6.1.2. Social and Technical Barriers

Participants acknowledged a number of social and technical barriers to explain why the dark data problem exists. Some barriers and solutions are addressed in subsections of the Discussion presented in this paper. Three commonly referenced barriers encapsulated in the following comment are: lack of funding, lack of time, and lack of credit.

Lack of interest in dark data by funding agencies. Lack of credit for curation of dark data for employment selection, promotion, or tenure. Lack of time for people on soft money to curate dark data. (Anonymous Respondent, 2016 pre-workshop survey).

Some of the proposed solutions relevant to this effort include development of centralized curation resources where costs could be amortized across projects and time, with skilled expert staff who can apply similar solutions across many projects and mechanisms to assure academic credit, including embargoes, data creator controls to access, and creation of DOIs for data:

The key issue is funding for initial ingestion and perpetual curation. There is also a social issue associated with private observatories who may need to be convinced to release their proprietary data products; funding could help here, too. Also, even major observatories have stores of dark data, e.g., the NOAO "save-the-bits" tapes. Interacting with these stakeholders will perhaps be more productive than scavenging piecemeal data sets from individual investigators. (Anonymous Respondent, 2016 pre-workshop survey).

6.1.3. Take Advantage of CyVerse Infrastructure and Longevity of University of Arizona

Leveraging dedicated CyVerse resources to support astronomical research and associated software requires identifying and incorporating corresponding software and metadata for astronomical data sets. Through CyVerse, it will also be possible to study user participation through established metrics, as well as observing broader applications for the astronomy and astrophysics communities as Astrolabe develops. Astrolabe will take advantage of both the CyVerse infrastructure as a computing environment and the astronomy activities on campus, as well as the longevity and stability of the University of Arizona as an institution, to establish a secure and robust repository that focuses on the needs of its users. Overall, Astrolabe shall support user control of assets and open access to research data, with a commitment to sustainability and education, in order to successfully navigate and avoid the following data curation challenges:

Inadequate human resources to process and document [data]. Loss of institutional memory of data properties. Lack of academic or professional credit for work on dark data that won't yield peer-reviewed publications. Unreadable legacy media. Scattered locations of the data. Loss of software needed to make sense of old

data. (Anonymous Respondent, 2015 pre-workshop survey).

However, as indicated by Sands et al. (2014), disciplinary expertise and buy-in are essential to support curation efforts. As a result, the interdisciplinary Astrolabe project team should be comprised of individuals familiar with techniques for astronomy data mining and visualization, as well as existing resources typically utilized by astronomers.

6.1.4. Obtain Community Buy-in and Manage Expectations

To solicit buy-in and to ensure that the systems being built meet community needs, an advisory board is critical for the project. Some board members may embrace active roles helping to incorporate their own data sets and advise project staff. Astrolabe must participate in a complex data space that includes many ongoing and completed observation missions, as well as data service projects and virtual observatories, where:

The community doesn't acknowledge or reward archiving your data except for a few large catalogs (and even those aren't rewarded enough) ... There aren't standards or tools for how to get your data into a archivable form that are easily accessible to the average small-team researcher. There aren't tools that allow other people to access data or even published tables quickly and easily. The VO effort spent a lot of time on standards, and basically produced things that are only of interest to large projects because the burden of making your data set or tool compliant with a VO standard is more than any individual has time to deal with, especially when they aren't getting paid to do it. (Anonymous Respondent, 2016 pre-workshop survey).

As envisioned, the Astrolabe project is essentially attempting to perform cultural engineering, changing the publishing habits of scientists to meet new open access requirements:

Technical issues are generally solved or solvable, but include data compression, accurate timestamps, attaching spatial and spectral coordinates, getting the community to cooperate on common standards, adopting underlying technologies likely to persist indefinitely, etc. (Anonymous Respondent, 2016 pre-workshop survey).

Chief resources to coordinate with include: the Harvard-Smithsonian Center for Astrophysics, ADS, and other repositories such as Dataverse and Zenodo. Furthermore, the UAT can be used in Astrolabe, providing a taxonomy for semantic enrichment.²³ In order to connect to existing resources such as ADS, it is necessary to establish a presence at key conferences and to meet with representatives of related projects, with the goals of educating the broader community about the new CyVerse resources and obtaining feedback from the community of users about their desires for the project's next steps through extramurally funded development. Finally,

²³ <http://astrothesaurus.org/>

education is a critical element for the viability and long-term sustainability of Astrolabe. This includes two main populations: astronomers who deposit data, and students in astronomy and information science who need to learn data science techniques. The Astrolabe team must work with authors of data to develop best practice guidelines for astronomy data, as well as to ensure that the data are in standard format and have appropriate metadata.

6.1.5. Focus on Low-hanging Fruit

Workshop participants identified two obvious pieces of “low-hanging fruit” as a niche opportunity for Astrolabe to provide valuable services to the astronomical community: dark data and other orphan data sets not curated elsewhere, and data associated with authors of articles published in AAS journals:

As a point of discussion, the highest priority could be assigned to the data underlying referred publications, along with medium-size data sets with uniform calibration that do not currently go to specific ground-based archives. (Anonymous Respondent, 2015 pre-workshop survey.)

As noted by Henneken & Accomazzi (2011), publications based on data sets are essentially expressions of data. Journal publishers are innovating new methods of digital publication that provide rich scientific data beneath a text publication itself. Furthermore, as citation rates appear to be higher for publications that contain links to referenced data, participating in data citation and persistent linking is an important objective of Astrolabe (Accomazzi 2011; Accomazzi & Dave 2011). Additional synergistic activities include working with the WWT,²⁴ a community-based tool for research, publishing, education, and public outreach. CyVerse represents an outstanding partner, and the CyVerse software stacks are very similar to others in use across the physical sciences community. Additionally, the AAS is working to develop a community-based software discovery portal for astronomy with robust developer workflows, unique identifiers, software citation, search, and developer credit (Henneken et al. 2017). Astrolabe should pilot an instance of this discovery portal in the data repository. Astrolabe is not limited to data that can be represented in WWT; it can store any data format for visualization, with packages such as Glue (Beaumont et al. 2015).

6.1.6. Develop a Follow-on Workshop

Workshop participants advised holding a subsequent workshop to connect community members as the system develops, to include both an education component and continued assessment of community needs. This follow-on workshop was held in 2016 and is described below. Additional future workshops should include data carpentry²⁵ plus hackathons, particularly targeting researchers with relevant data sets.

6.2. 2016 Astrolabe Workshop Outcomes

2016 Astrolabe follow-on workshop participants indicated the following important areas of focus, which are now priorities for ongoing system development.

6.2.1. Data Complexity, Heterogeneity, and Physical Format of Dark Data

It is rare that literally retracing someone else’s data analysis path is going to be worthwhile. What we have is the issue of reusability. Our data should often be useful for multiple projects, but access problems make that more difficult than it needs to be. (Anonymous Respondent, 2016 pre-workshop survey).

Historical data can be made accessible through Astrolabe. Data sets stored on magnetic tapes at NOAO and the Lunar and Planetary Laboratory (LPL) would likely be of interest to the community. Promising funding sources for these efforts were identified by UA Library participants, specifically for data preservation in the U.S. national library space. Similar efforts at the Harvard-Smithsonian Center for Astrophysics involving astronomical plates represent a possible partnership for archival expertise,²⁶ and additional funding opportunities exist for such efforts. These data represent an ideal opportunity for citizen science projects as well.

Larger data sets can also fall into the category of dark data, with complexity of the data leading to inaccessibility for reuse and research transparency:

There are also very complex data sets that are difficult to host, think mini-SDSS type databases that include spectra, images, figures and tables all integrated by a search engine. How do you integrate something like that? (Anonymous Respondent, 2015 pre-workshop survey).

Furthermore, individual missions are able to specify the format of their data to create a certain amount of consistency. Individual researchers, however, structure their data to individual needs:

Astronomers don’t want to properly separate final data products from intermediate files in their “working directory.” A final measurement may come from a file having some unique pattern in the name, but is mixed in with all the files leading to the creation of that spectrum and only the astronomer understands the naming scheme. (Anonymous Respondent, 2015 pre-workshop survey).

This means that a repository designed to support dark data must have staff and cyberinfrastructure that can manage this difficulty without putting undue burden on the individual researchers:

²⁴ <http://worldwidetelescope.org>

²⁵ <http://www.datacarpentry.org/>

²⁶ <https://platestacks.cfa.harvard.edu/>

How to make such a heterogeneous set of data searchable? Extractable? (Anonymous Respondent, 2015 pre-workshop survey.)

If cyberinfrastructure is shared across projects, it can be more cost effective and long-lived than a scenario in which each researcher builds infrastructure for their own instruments or projects. It is also fundamentally important that any new service dealing with uncurated data should coordinate with other repositories, in order to maximize resources and expertise, as well as avoid redundant efforts:

There are other efforts to create a data hub on a national scale. I am aware of a group at the National Center for Supercomputing Applications and the National Data Service who are working to build a massive data hub for storage of data sets pan-science. To date, they have a working prototype involving manipulation of data in situ on their servers to avoid too many large data transfers of massive data sets. It may be worthwhile to communicate with them to see if a shared infrastructure is warranted. Two autonomous efforts by the both of us may not be as strong as one where we work together. (Anonymous Respondent, 2015 pre-workshop survey).

6.2.2. Author Websites Archiving Data

Authors frequently link to data on personal websites from the literature, and journals do not enforce a guarantee that these websites will be long-lived:

Often, collaborations that are medium size (multi year but much smaller than SDSS) create databases to serve their own uses and the public. These generally fail some years after the collaboration is done, because the database software gets out of date or a server is upgraded, etc., and no one is getting paid any longer to make sure it's available. Maybe this is an application for the proposed data hub ... I think you would do a great service if you could make published data easily usable and cross-referenceable. Dark data is interesting but re-extracting quantities from it is unattractive, especially if they have already been published. (Anonymous Respondent, 2016 pre-workshop survey).

Astrolabe could provide authors with a website tool through CyVerse, ensuring the integrity of links in published literature. This would allow authors to freeze a site at publication to preserve a record of the research process. Several existing web pages should be ingested as a proof of concept. Search tools are necessary to locate files across collections. Templates could be created through CyVerse, and users could adopt templates for the appropriate data structure, including key presentation and descriptive information. Metrics are needed to demonstrate usage.

6.2.3. Time Domain and Serendipitous Data Cases

A key opportunity for Astrolabe is supporting time domain astronomy, including follow-up to Large Synoptic Survey Telescope (LSST) observations, and bringing together data sets and researchers from multiple telescopes, if possible:

Re-observation of a very faint target can be expensive in telescope time. Missing a point in a time domain sequence can impact the interpretation, or require negotiation with a person that the investigator did not necessarily intend to be a co-author. (Anonymous Respondent, 2015 pre-workshop survey).

Darch & Sands (2017) show that the LSST project faces substantial uncertainty regarding future stakeholder requirements, including users' research practices, required tools and expertise, and the ability of other instruments to join the network for follow-up observations. Open source development places a burden on the community to create adaptive infrastructures to effectively interface with LSST data. Digital libraries such as Astrolabe must therefore anticipate and respond to change, promote standardization, develop infrastructures that empower users to adapt and reconfigure as needed, and adopt open source policies.

Another potential source of serendipitous discovery is historical data on nearly obsolete, institutional, and individual removable media (as discussed above), which could be converted to current formats and ingested into Astrolabe. Historical data represent an opportunity for citizen science and educational opportunities. The amateur astronomy community is also a potential target for data. It is recommended that Astrolabe start with a minimum of 100 terabytes of storage space, planning to scale up to 10 petabytes over 10 years. The system should include the following features: support for a visualization interface; minting DOIs; checksum to detect errors in file transfer; licensing; capability to upload notes and other historical metadata related to data; public/private options and related policies; support for building tools; and capturing provenance of data. All of these features can be supported by CyVerse technology. To build a sustainable framework, Astrolabe should consider federation with other systems, and highlight costs saved by utilizing CyVerse in proposals to NSF and other agencies, including private foundations.

6.2.4. Searching the Literature for Dark Data

Text mining could locate references to dark or potentially dark data in the literature. Challenges associated with curation of these data provoke the following questions:

How do you make "dark data" findable? How to improve the "citability" of "dark data"? Does it make sense to write "data release" papers for "dark data," which would improve findability in publication-centric discovery services like the ADS? How do you measure the impact of "dark data" or at least whether it is being used? (Anonymous Respondent, 2016 pre-workshop survey).

With the availability of a convenient resource for archival such as Astrolabe, authors could be incentivized to share data. High-impact papers and publications within the past 10 years could be targeted first. The ADS could help with relevant expertise and community support, as well as text mining to locate papers without data links. Candidate data associated with an existing publication would advertise for Astrolabe and provide growth and overall sustainability for the project.

The standards of the Virtual Observatory support searching across repositories, but also lead to challenges associated with data heterogeneity:

Contrast with the Virtual Observatory paradigm that cooperates on standards but has no central structure to ensure continuity. (Anonymous Respondent, 2016 pre-workshop survey).

7. Discussion

Workshop and survey participants overwhelmingly confirmed that dark and legacy data—along with some data associated with current small- to medium-sized projects—in astronomy represent a valuable target for curation efforts. The issues associated with dark or at-risk data can be organized into three categories, which were discussed at length in both Astrolabe workshops, and are summarized here: (1) the science use cases for bringing dark data to light; (2) where to find useful dark data; and (3) governance, sustainability, and buy-in.

7.1. Science Use Cases

Any repository needs to support new science as a core justification for its existence. While it is impossible to determine what data will be valuable in the future, there are indications of what might be most valuable. As discussed extensively in Section 6 above, Astrolabe is scientifically justified and should increase communication of prior work for integration with new data, reanalysis under new methods with new data, verifiability, and replicability. One example is the ability to revisit prior observations when a celestial event is recorded with LSST, which was highlighted in Astrolabe workshops as a strong motivator to look back in time to old observations. LSST will allow researchers to identify events across large sections of the sky, rekindling interest in how objects of study were behaving in the past. Availability of historical data would contribute to ongoing study of phenomena that change over time. Proper motion benchmarks represent another important application for Astrolabe archival data.

7.2. Finding Dark Data

Many workshop participants and Astrolabe Advisory Board members have provided pointers for locating valuable data in danger of being lost. This includes not only older data, but also current or recent projects with uncertain sustainability. Three main types of old data sources were identified. The first is large data collections that were intentionally created for preservation purposes but now are underutilized because of data format obsolescence. The second is data associated with publications. The third is data from instruments that are no longer operating. Some of these data may have an institutional home, but because of time and costs, have not been migrated to more accessible formats and platforms. One of many examples is NOAO’s tape

data archive. We are currently developing plans to move 100 terabytes of exabyte tapes and other media to more accessible formats. Our workshop participants have further identified many facilities and individuals with boxes of media of different types in storage rooms and filing cabinets worldwide. Previous work of this type includes the Digital Access to a Sky Century @ Harvard (DASCH) project that aims to scan the majority of the Astronomical Photographic Plate Collection’s 500,000 glass plate negatives and produce full photometry results for the entire sky (Simcoe et al. 2006).

Many publishers are now making some data associated with journal articles available online. However, this is not universal and few such records exist for older publications. Because not all historical data can be collected, participants suggested focusing on: high-impact articles under the reasonable assumption that these data would be of greatest use; recent publications because the probability of recovering the data from the authors is highest; and finally, identification of outdated or broken data references and links in existing publications. While the three categories described here only address a small portion of the data that participants identified as important, Astrolabe is working with the AAS in each of these areas. We are currently searching the last 10 years of publications for interesting data, with plans in place to create semi-automated methods of identifying data references, beginning with locating broken hyperlinks in publications. There are also linguistic references to data that do not always refer to existing collections, but these are more difficult to identify in large collections of literature. For example, a paper may refer to a data set from a particular observing instrument that was operational at the time of publication, but the instrument’s data archive is no longer available online. We plan to develop pattern matching and machine learning tools to help identify these references.

Current projects are also sources of dark data, and this category overlaps substantially with new funding models. We are unable to prevent new data from going dark unless we solve the financial problems associated with storing, processing, and disseminating data. The same social and technical influences that cause older data to go dark can also impact newer data. For example, three-year agency-funded projects can produce data streams that are difficult to maintain after the funding period expires. The widespread requirements for data management plans are not accompanied by additional funding for data management beyond the duration of a project. Smaller instruments and missions may not be archived in a long-lived repository, even if all data collected are not analyzed by the end of the funded project. Astrolabe is developing partnerships and new research proposals with current instrumentation projects to construct efficient workflows for new data and digitization of old data that are identified as potentially useful.

7.3. Governance, Sustainability, and Buy-in

Finally, the community has pointed out the critical element of governance, sustainability, and buy-in. We have formed an advisory board of experienced community members to oversee development and funding. We have identified several principles that capture some of the important points discussed above. The Astrolabe repository must be open to all researchers, facilitating efficient access to data. It must have a low cost of operation. Wherever possible, it must reuse or improve appropriate technologies, such as the VO initiative and pan-science initiatives

like Cyverse and JetStream, rather than build new technologies. It must be able to provide services that will encourage new projects to buy-in to meet their own goals and assure the long-term operation of facilities. The repository must be more than a passive storage location for data. Astrolabe needs community involvement to install and process data and metadata. Each of the two workshops presented here directed the Astrolabe project team toward key uncurated or at-risk data sets and strategies for locating and curating these data. Some of the ideas can be incorporated into existing repositories and many are being built into the Astrolabe repository.

One potential strategy for sustainability is to pull together resources from many data-gathering projects staggered over time. Another partial solution can be found from an analogy to publication of scholarly papers. A percentage of the overhead costs from research grants received by university researchers is designated for supporting institutional journal subscriptions, which then supports the editing and publishing of journals. Libraries maintain digital and paper collections also supported in part by overhead on research grants. Data management might in part be supported by overhead on grants, but this model for peer-reviewed publication has resulted in economic problems with rising journal subscription costs making access prohibitively expensive for individuals and institutions. The open access movement now directs some of the costs away from subscriptions, and this model might also be applied to data with a data deposit fee. Such a fee could be built into grant proposals, but ultimately would need to cover the cost for curation throughout the entire data lifecycle. If Moore's law holds for data storage, perhaps some current data costs can be covered with future projects that would require much larger data sets. The Dryad repository for life sciences is an example of a successful business model for nonprofit governance through a membership organization.²⁷

8. Conclusions and Future Directions

Data creators and researchers typically require motivation and incentive to share data, such as a clear vision for future use of the data in a broader context and professional credit for reuse. There must be low barriers to the technical process of actually putting data into a repository. Furthermore, researchers must maintain a certain level of control of their data over time to ensure proper attribution and quality control. We identified factors that impact the willingness and ability to share and use scientific data in astronomy, considering that data are only as useful as the software used to analyze and visualize the data. Beyond simply creating another data repository, participants in the two workshops discussed here identified advantages to using cyberinfrastructure that could facilitate not only data deposition and subsequent discovery, but also significant processing of that data, resulting in the creation of Astrolabe as a community resource that is now available for widespread use by astronomers through CyVerse.

New instrumentation and accelerated data flows in astronomy are posing a challenge to traditional approaches to experimentation and computing. For small- and medium-sized projects, there are no clear options for long-term sustainability in new data management plan requirements (Hanisch et al. 2017). CyVerse and related platforms such as JetStream²⁸ (Towns et al. 2014; Stewart et al. 2015) exemplify the type of

beneficial curation and computational environment envisioned by workshop participants. This cloud-based approach allows for flexible computing where one to hundreds of processors can be applied to a problem. We will incorporate a community-developed controlled vocabulary (i.e., the UAT) to enhance findability of data sets. These data will be processed in CyVerse and then integrated with the WorldWide Telescope visualization tool by the end of our current National Science Foundation grant in 2019. Further supporting discoverability, we have developed capabilities for FITS header extraction and semi-automated population of metadata templates configured according to the IVOA ObsCore, FITS, and WCS standards. Furthermore, when data sets in Astrolabe are determined to be static and ready for public access, CyVerse is capable of minting DOIs with a DataCite²⁹ template, which permits linking to Astrolabe from ADS and connecting data to literature.

This paper has described the data curation recommendations of participants in two community workshops, and we have provided details about the ongoing development of the Astrolabe repository system to enable curation of potentially valuable dark or at-risk data in astronomy. Overall, we wish to expand the capacity of the system to increase ease of use, data discoverability, convenient access to data manipulation, and incorporating additional data types, including not only data being generated from instruments aimed into space, but also remote sensing instruments pointed back down toward earth, such as Organization of Biological Field Stations site data that could also be visualized in WWT. Shared infrastructure presents curation challenges that must be navigated with the participation of users. Establishing an active Astrolabe user community will help us to identify both domain-specific computational tools and data for the shared environment, and also to exploit domain-spanning cyberinfrastructure, enabling grand challenge research and serendipitous discovery in multiple domains.

We gratefully acknowledge the support of the National Science Foundation (grant number 1642446, P.B. Heidorn, PI; J. Steffen, Co-PI), the University of Arizona (2015 "Start for Success" grant and 2016 "Accelerate for Success" grants, P.B. Heidorn, PI; D. Zaritsky, Co-PI), the American Astronomical Society, and CyVerse. We thank our additional collaborators at the University of Arizona, including the UA Department of Astronomy and the University Libraries, as well as workshop organizers and participants for their time, engagement, and expertise. We especially thank Phil Rosenfield for his thorough feedback on the manuscript, along with our anonymous reviewer.

ORCID iDs

P. Bryan Heidorn  <https://orcid.org/0000-0002-4601-8180>
 Gretchen R. Stahlman  <https://orcid.org/0000-0001-8814-863X>
 Julie Steffen  <https://orcid.org/0000-0002-8596-6634>

References

Accomazzi, A. 2011, Future professional communication in astronomy II (Cambridge: FPCA-II), 135
 Accomazzi, A., & Dave, R. 2011, Semantic interlinking of resources in the virtual observatory era, arXiv:1103.5958

²⁷ datadryad.org

²⁸ <https://jetstream-cloud.org/>

²⁹ <https://www.datacite.org/>

Accomazzi, A., Kurtz, M. J., Henneken, E. A., et al. 2016, Aggregation and Linking of Observational Metadata in the ADS, arXiv:1601.07858

Anderson, C. 2004, *Wired*, **12**, 170

Atkins, D. 2003, Revolutionizing science and engineering through cyberinfrastructure: Report of the National Science Foundation blue-ribbon advisory panel on cyberinfrastructure (<https://www.nsf.gov/cise/sci/reports/atkins.pdf>)

Beaumont, C., Goodman, A., & Greenfield, P. 2015, Astronomical Data Analysis Software an Systems XXIV (ADASS XXIV) (Sydney: ADASS), **101**

Borgman, C. L. 2015, Big Data, Little Data, No Data: Scholarship in the Networked World (Cambridge, MA: MIT Press)

Brunner, R. J., Djorgovski, S. G., & Szalay, A. S. 2001, Virtual Observatories of the Future (Pasadena, CA: ASP)

Choudhury, S., DiLauro, T., Szalay, A., et al. 2008, *International Journal of Digital Curation*, **2**, 20

Conrad, A. S., Rasmus, H., & Svendsen, M. 2017, *International Journal of Digital Curation*, **12**, 37

Cragin, M. H., Palmer, C. L., Carlson, J. R., et al. 2010, *RSPTA*, **368**, 4023

CyVerse 2017, Transforming science through data-driven discovery: CyVerse services overview, 2017 (<https://pods.iplantcollaborative.org/wiki/download/attachments/29823820/CyVerse%20Services%20Overview%20Brochure%202017%20updated.pdf?version=1&modificationDate=1486747053412&api=v>)

Darch, P. T., & Sands, A. E. 2017, in Joint Conference on Digital Libraries, 1

Edwards, P. N., Mayernik, M. S., Batcheller, A. L., et al. 2011, *SSS*, **41**, 667

Folk, M., Heber, G., Kozioł, Q., et al. 2011, in Proc. EDBT/ICDT 2011 Workshop on Array Databases (Uppsala: ACM), 36

Frey, K., & Accomazzi, A. 2018, *ApJS*, **236**, 24

Greenfield, P., Droettboom, M., & Bray, E. 2015, *A&C*, **12**, 240

Grosbol, P. 1988, *BICDS*, **35**, 7

Hanisch, R., Hildreth, M., McEwen, L., et al. 2017, MPS Open Data Report (<https://curate.nd.edu/show/hm50tq60k56>)

Hanisch, R., Steffen, J., Choudhury, S., et al. 2007, in Library and Information Services in Astronomy V (Cambridge: Wolbach Library), **29**

Heidorn, P. B. 2008, *Library Trends*, **57**, 280

Henneken, E. 2015, *Bulletin of the Association for Information Science and Technology*, **41**, 40

Henneken, E., Accomazzi, A., Blanco-Cuaresma, S., et al. 2017, Asclepias—Capturing Software Citations in Astronomy, Zenodo, <http://doi.org/10.5281/zenodo.1011088>

Henneken, E., Accomazzi, A., Kurtz, M., et al. 2009, Astronomical Data Analysis Software and Systems XVIII (Québec City: ADASS), **384**

Henneken, E. A., & Accomazzi, A. 2011, arXiv:1111.3618

Keeney, S., Hasson, F., & McKenna, H. P. 2001, *Int. J. Nurs. Stud.*, **38**, 195

Lenhardt, W. C., Conway, M., Scott, E., et al. 2016, in 2016 IEEE, High Performance Extreme Computing Conf. (HPEC) (Waltham: IEEE), 1

McCray, W. P. 2004, Giant Telescopes: Astronomical Ambition and the Promise of Technology (Cambridge, MA: Harvard Univ. Press)

McCray, W. P. 2014, *Technology and Culture*, **55**, 908

Michener, W. K., Brunt, J. W., Helly, J. J., et al. 1997, *Ecol. Appl.*, **7**, 330

National Academies of Sciences, Engineering, and Medicine 2016, Future Directions for NSF Advanced Computing Infrastructure to Support US Science and Engineering in 2017–2020 (Washington, DC: National Academies Press)

Novacecu, J., Peek, J. E. G., Weissman, S., et al. 2018, *ApJS*, **236**, 20

NSF 2015, Public Access Plan: Today’s Data, Tomorrow’s Discoveries: Increasing Access to the Results of Research Funded by the National Science Foundation (https://www.nsf.gov/publications/pub_summ.jsp?ods_key=nsf15052Report_number_is:nsf15052)

Open Science Collaboration 2012, *Perspectives on Psychological Science*, **7**, 657

Rosenfield, P. 2018, *ApJS*, **236**, 22

Sands, A. E., Borgman, C. L., Traweek, S., et al. 2014, *International Journal of Digital Curation*, **9**, 98

Simcoe, R., Grindlay, J., Los, E., et al. 2006, arXiv:astro-ph/0610351

Stahlman, G. S., Heidorn, P. B., & Steffen, J. 2018, Proceedings of Library and Information Services in Astronomy (LISA) VIII (Strasbourg: CDS), in press

Stewart, C. A., Cockerill, T. M., Foster, I., et al. 2015, Jetstream: A self-provisioned, scalable science and engineering cloud environment (<https://scholarworks.iu.edu/dspace/handle/2022/20286>)

Towns, J., Cockerill, T., Dahan, M., et al. 2014, *CSE*, **16**, 62

White, R., Accomazzi, A., Berriman, G., et al. 2009, Astro2010: The Astronomy and Astrophysics Decadal Survey, **64**