

Astronomical Data Archives
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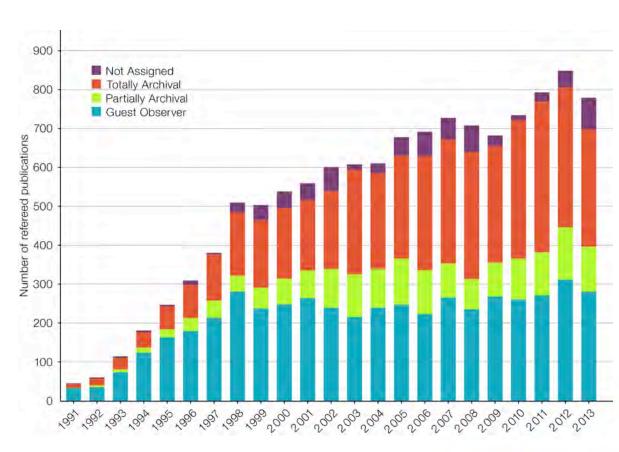
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- 1. Archive Goals
- 2. Current Archives
- 3. Current and Upcoming Challenges



## Astronomy Archive Goals

- Data safe-keeping
  - Tremendous financial and community resources are put into gathering data,
     so some effort in preserving data for the long run is justified
  - Example: Infrared Astronomical Satellite data (IRAS; launched in 1983) was still cited times more than 200 times in 2017
- PI access
- Archival research
  - For Hubble, Archival research (i.e., not from the proposing team, comprises ~half of all publications)





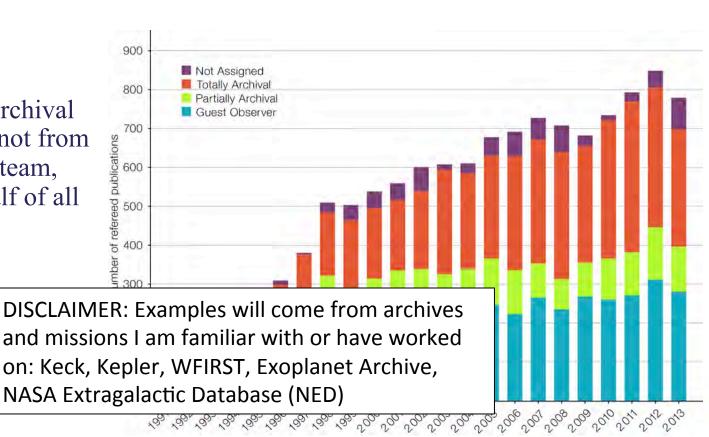
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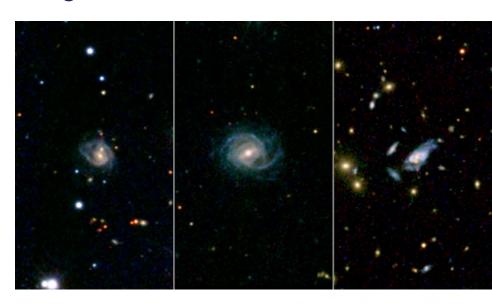
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## Archival Research Example

- Easily searchable data facilitates innovative uses
  - Example: Ogle et al discover super spiral galaxies using NED (NASA Extragalactic Database)
    - > These galaxies have luminosities 10x the Milky Way
  - No new data taken
  - $\circ$  Original goal was to search of properties of 800,000 galaxies (z < 0.3) to find luminous elliptical brightest cluster galaxies
  - Database combines information from
    - Galaxy Evolution Explorer (Galex)
    - Sloan Digital Sky Survey
    - > Two Micron All-Sky Survey (2MASS)
    - > Spitzer
    - > Wide-field Infrared Survey Explorer (WISE)







### **Current Archives**

- Archives contain:
  - o Raw and processed data from the mission/team
  - o High level data products from the community
  - o High level products/associations from the archive (e.g. cross identifications)
- Archive levels and support vary greatly, from non-searchable raw data to processed data with visualization and tools
- Congressional mandate on federally-funded data has accelerated the trend of PIproduced data being made available to the community
- Data release models
  - Continuous: most GO driven telescopes/missions, with or without proprietary period
  - Episodic: Generally when processing is needed on sets of data (spatial or temporal)

For recent updates and developments: Astronomical Data Analysis Software & Systems conference series



## Current Archives: General Categories

- Space missions
  - NASA, ESA, JAXA etc. generally provide high level data archives including processed data and often analysis tools
  - Funding to develop and operate the archive is generally allocated as part of the mission planning process
- Ground-based telescopes
  - o Optical
    - > Most large telescopes have archives (Keck, VLT/ESO, Gemini etc)
    - > Many small and medium sized telescopes do not have open archives
  - o Radio/millimeter
    - > NRAO, ALMA





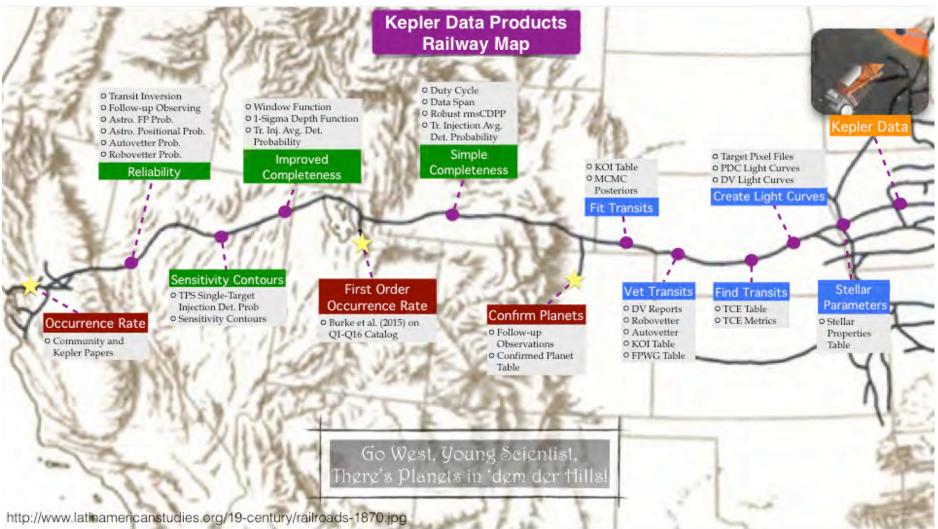
### Ancillary Data and Additional Products

- Need to understand what additional data is crucial for data processing or interpretation
  - For processing
    - > Spacecraft data (even if it doesn't seem immediately relevant)
      - \*Example: Spitzer developed an exoplanet transit mode ~8 years after launch using pointing data not originally used in processing pipeline
  - For interpretations/astrophysics
    - > Example: Source categorizations
    - > More likely to come from community, issues with standardization and reliability
    - > LISA example: Input waveforms
- Other kinds of data products
  - o Completeness and reliability or pipeline characterization data
  - Simulated data and inputs
  - Probabilities and posteriors





## Example: Kepler

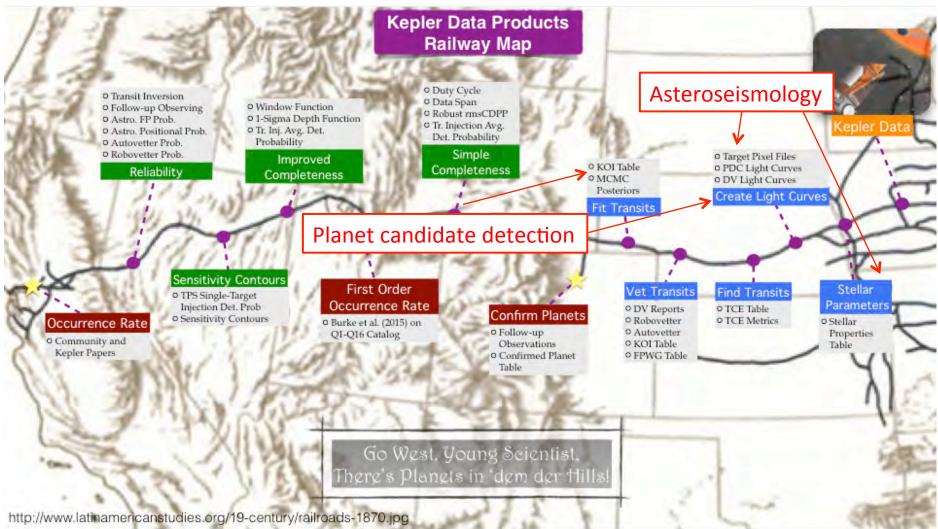




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## Example: Kepler



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# Current and Upcoming Challenges: Data Volume (1)

- By Earth Sciences or Silicon Valley standards, Astronomy is not "Big Data"
- BUT, data volume is still an issue
  - Current examples
    - > WISE all sky survey
      - ❖Source catalog ~ 800 million sources
      - ❖ Photometry measurements ~ 42 billion
    - > IPAC total holdings = 12 PB
  - Upcoming mission examples
    - > LSST max data rate = 20 TB/night (2020)
    - > WFIRST max data rate = 1.5 TB/day (2026)
    - > SKA (Square Kilometer Array) = 20 TB/day processed data (2024, partial)
- Issues with volume:
  - o Retrieval
  - o Organization/Searching
  - o Processing





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- Issues with volume:
  - ∘ Retrieval →
  - Organization/Search
  - Processing

Some users want (or think they want)

the entire survey. Adds significant

logistics and data management work.



## Data Volume: Organization/Searching

#### Lessons learned

- Simple solutions with careful planning can get you a long way without much technical magic: *organization* is the magic!
- Reduce s/w overheads: small latencies that didn't use to matter now stand out.
- Beware complexity things that are complicated when they are small explode on you when they grow big.
- Large datasets are hard to move: try to get it right the first time, and consider moves carefully.
- Large databases are hard to change or update, so plan the content carefully before loading.
- o Optimize data layouts for most common use cases: **But** different use cases require different organizations.
- May need to consider indexing in space-time, rather than just space, for moving object applications.



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## Data Volume: Processing

- Large data volumes drive the need to process the data where they are stored
  - o Not all users will have the resources to store and/or process large volumes
  - Also an issue for heavy processing on mid-sized data
- Different models
  - Give users CPU resources on same system as archive, generally comes with a version of the pipeline which may may editable by user
  - o Store copy of the data somewhere with large processing capabilities
    - > Cloud
    - > Super computer center
  - o Users often want to start at intermediate processing level
- Lesson learned: Best if planned and developed with rest of telescope
  - Even if not funded immediately, having a design minimizes the chance that later capabilities will be precluded or cost much more to develop





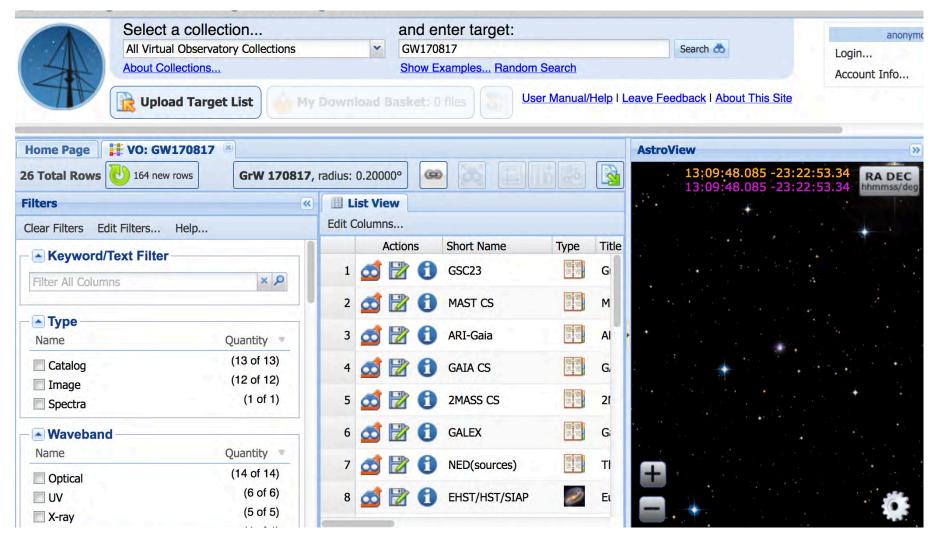
## Current and Upcoming Challenges: Multi-mission Research and Interoperability

- Many areas of research require coming several wavelengths/mission/ measurement types to make progress on the astrophysics
- Given the many different sources of archive funding (and different mandates/requirements) it is unlikely that any group would be funded to store ALL astronomy data
- Virtual Observatory efforts
  - IVOA and national VO organizations have worked to set standards and provide tools and services to allow users to access data from multiple host archives
  - Defining data standards gets progressively harder as data types become more complicated, e.g. data tables vs spectral cubes from an integral field spectrograph
  - o Implementation within archive done with resources from individual archives
- Lesson learned: Need to design good interfaces for users to retrieve data with complex constraints as science cases will always be incomplete





## VO Example: GW170817



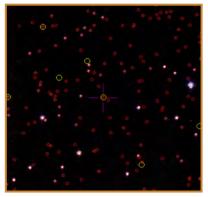


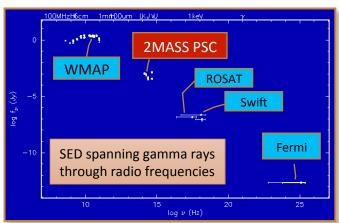
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# Current and Upcoming Challenges: Cross Matching/Identification

- One of the most important functions of an archive is matching sources between catalogs
  - Many different levels
    - > Coordinate only
    - > Coordinate + astrophysics
    - > Measure of reliability



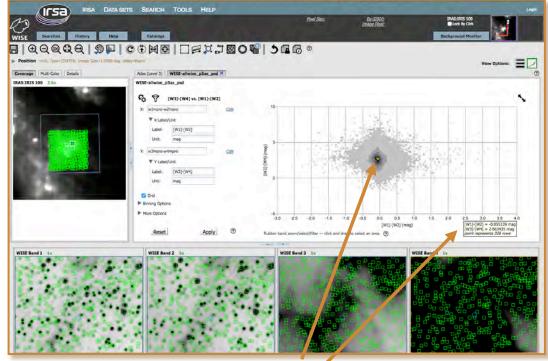


- Example:
  - NED: (NASA Extragalactic Database)
- Must correlate newly ingested catalogs with existing database of ~250 million objects.
- Currently ingesting 2MASS catalog with 470 million sources.
- Machine learning used for assigning statistical probabilities to matches



## Current and Upcoming Challenges: Data Visualization

- Interactive graphics provide intuition about the data.
- Co-registration of data sets: Example: IRSA allows simultaneous viewing of different data sets.
- Time-domain: light curves, folded-viewing, periodograms, moving objects.
- For massive sets we have to go from symbol representation to continuous quantities: density plots, histograms.
- Data Cubes



IRSA Viewer uses a density plot when the number of points becomes too great to show individually. The number of points in each bin in the plot is provided on hover.

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### What does LISA need?

- Is the LISA catalog a table with the same parameters for all sources/detections?
- Consider hierarchical or on-demand products
  - o Fermi
    - > User driven generation of products based on current data
    - > See info from Anne on Google drive
  - Kepler
    - > Different products for different use cases
    - > Additional/ancillary data varies depending on object
- LISA can use a combination of previous and current formats and structures and custom ones
- Beware of overly complex products
  - o i.e. if a table is mostly empty, it may be more useful as multiple products
- High level archive design considerations
  - Collection of products at each processing level vs single series of products
  - What is generated by project/science team and what is contributed by community





## Takeaway Points

- A well designed archive enhances what the community can do
  - Design depends on both on raw and processed data types, but also the most common use cases
- There are many different archive types and organizational models to choose from
- Be optimistic in archive plan but be prepared for reality
  - o Include many use cases, but prioritize
  - o Plan for future expansion
  - Even if processing development has to be delayed, put effort into design and use cases
- Use whatever you can from the community
  - o It's nice to think about doing "everything right" but resources are always finite

