



Astro Data Lab

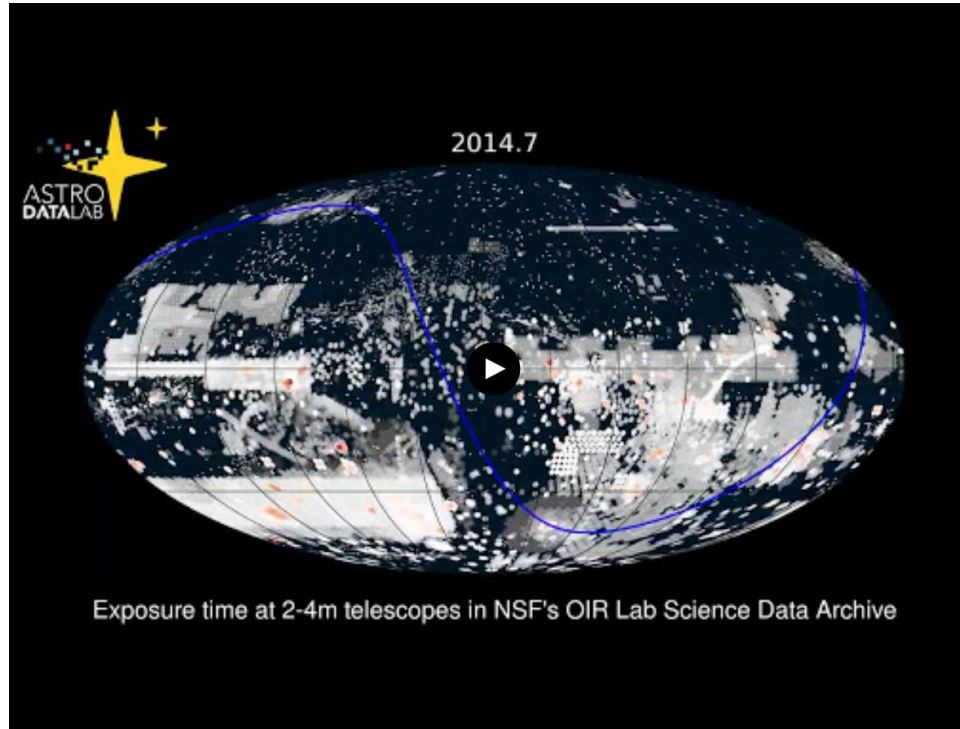
An Open-Access and Open-Data
Science Platform

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NOIRLab

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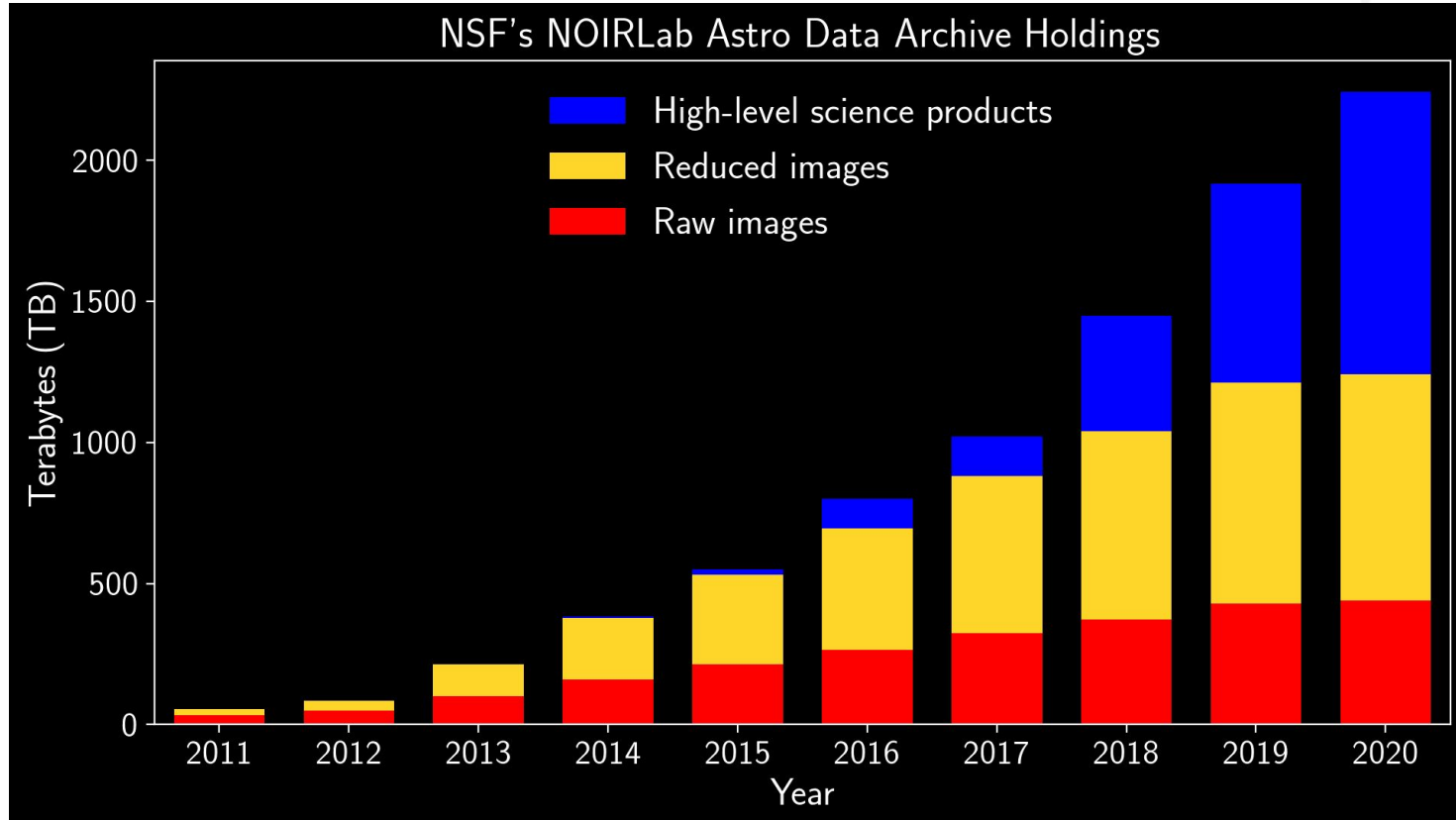
Motivation - The Data Avalanche



Exposure time at 2-4m telescopes in NSF's OIR Lab Science Data Archive

<https://youtu.be/lbRWdOqWrEk>

Data growth





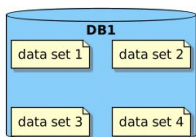
From data set to Science Platform



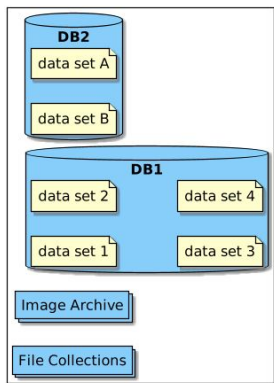
1
Data Set



2
Database

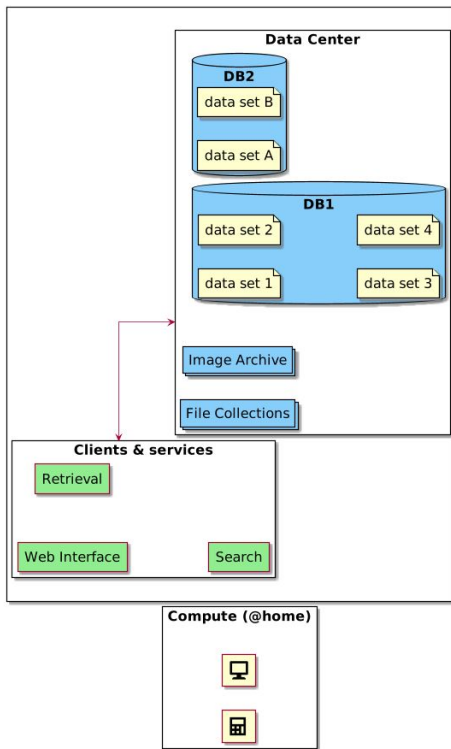


3
Data Center



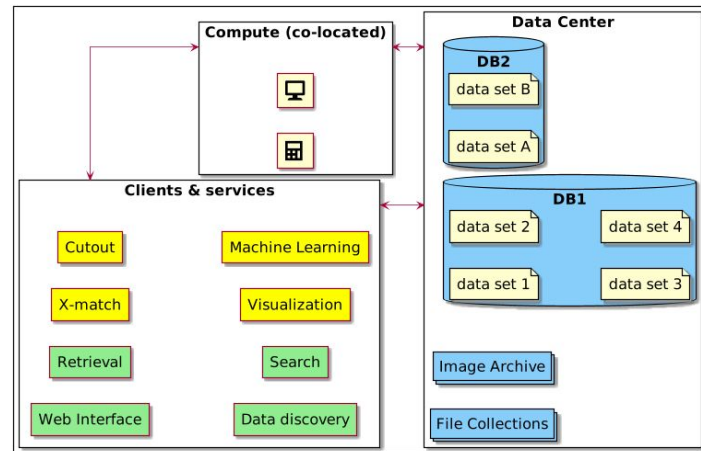
4

Science Archive



5

Science Platform



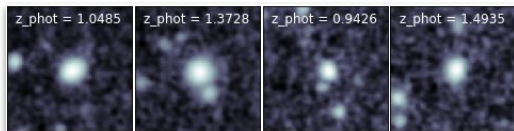
Data types at Data Lab

Currently

Catalogs
(2D tables)

	ls_id	ra	dec	dereid_mag_r	dereid_mag_g
0	8797229232750724	286.604936	43.783519	19.3421	20.7393
1	8797229232750718	286.602226	43.780599	22.8721	23.0592
2	8797229232750733	286.603586	43.786786	22.9804	23.4134
3	8797229232750742	286.612393	43.790177	18.8789	20.2648
4	8797229232750743	286.612561	43.791592	20.5371	22.0037
5	8797229232750735	286.607780	43.788338	19.2442	19.7413

Images
(2D arrays)

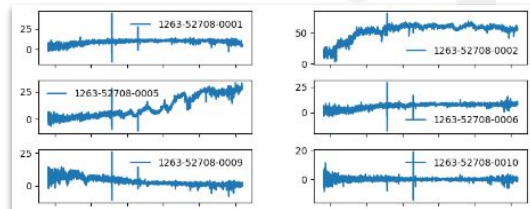


Heterogeneous
data collections
(file service)

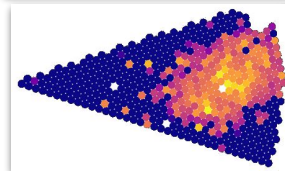
```
1 | print(sc.ls('gogreen_dr1://', format='long'))
drw-rw-rw- gogreen_dr1 0 13 Aug 2020 17:54 CATS
drw-rw-rw- gogreen_dr1 0 13 Aug 2020 17:54 PHOTOMETRY
-rw-rw-rw- gogreen_dr1 5429 13 Aug 2020 17:54 README
drw-rw-rw- gogreen_dr1 0 13 Aug 2020 17:54 SPECTROSCOPY
drw-rw-rw- gogreen_dr1 0 13 Aug 2020 17:54 Scripts
```

Currently

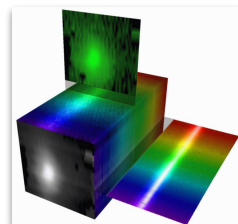
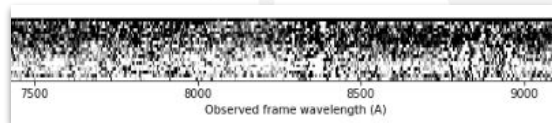
1D spectra
(queryable)



Soon
DESI



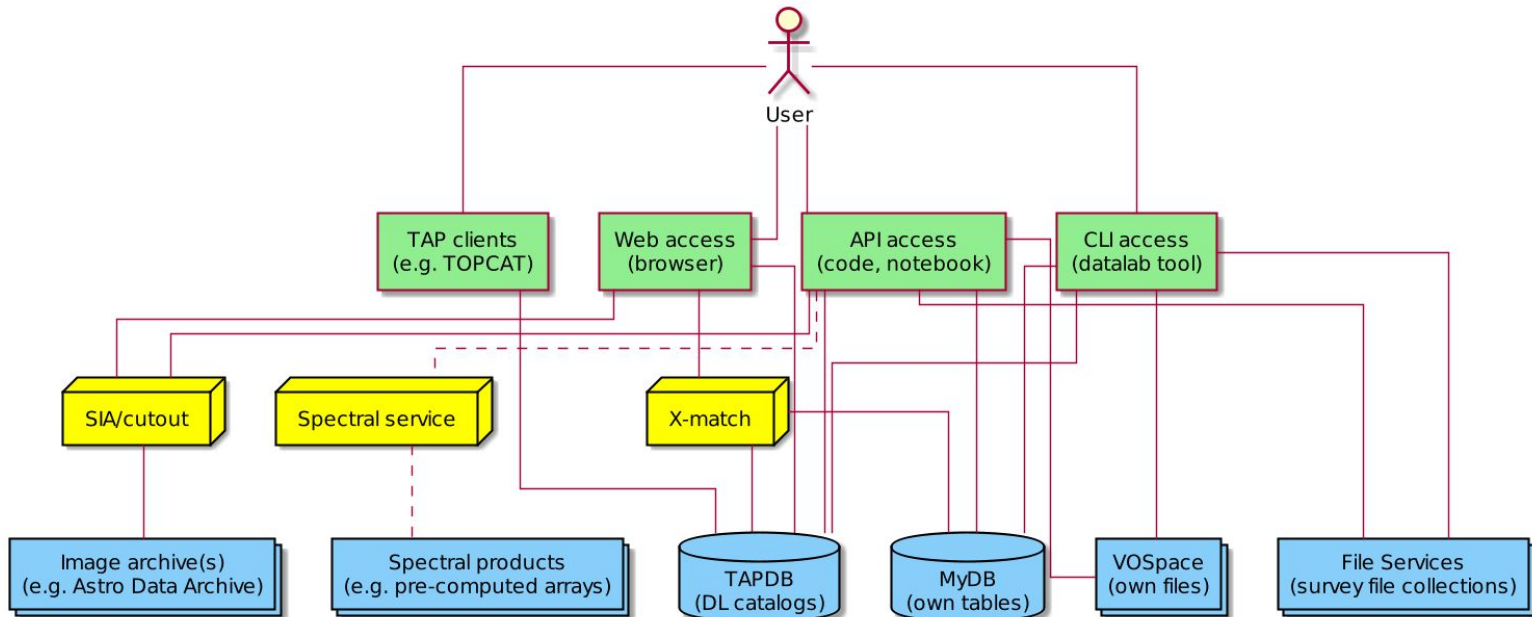
Future
2D spectra
IFU cubes &
complex data



GMOS-IFU,
MaNGA,
US-ELTs,
...

Scientific services at Data Lab

Data Lab services from a user's POV



Use these front-ends...

...to interface these services...

...to/or access these data

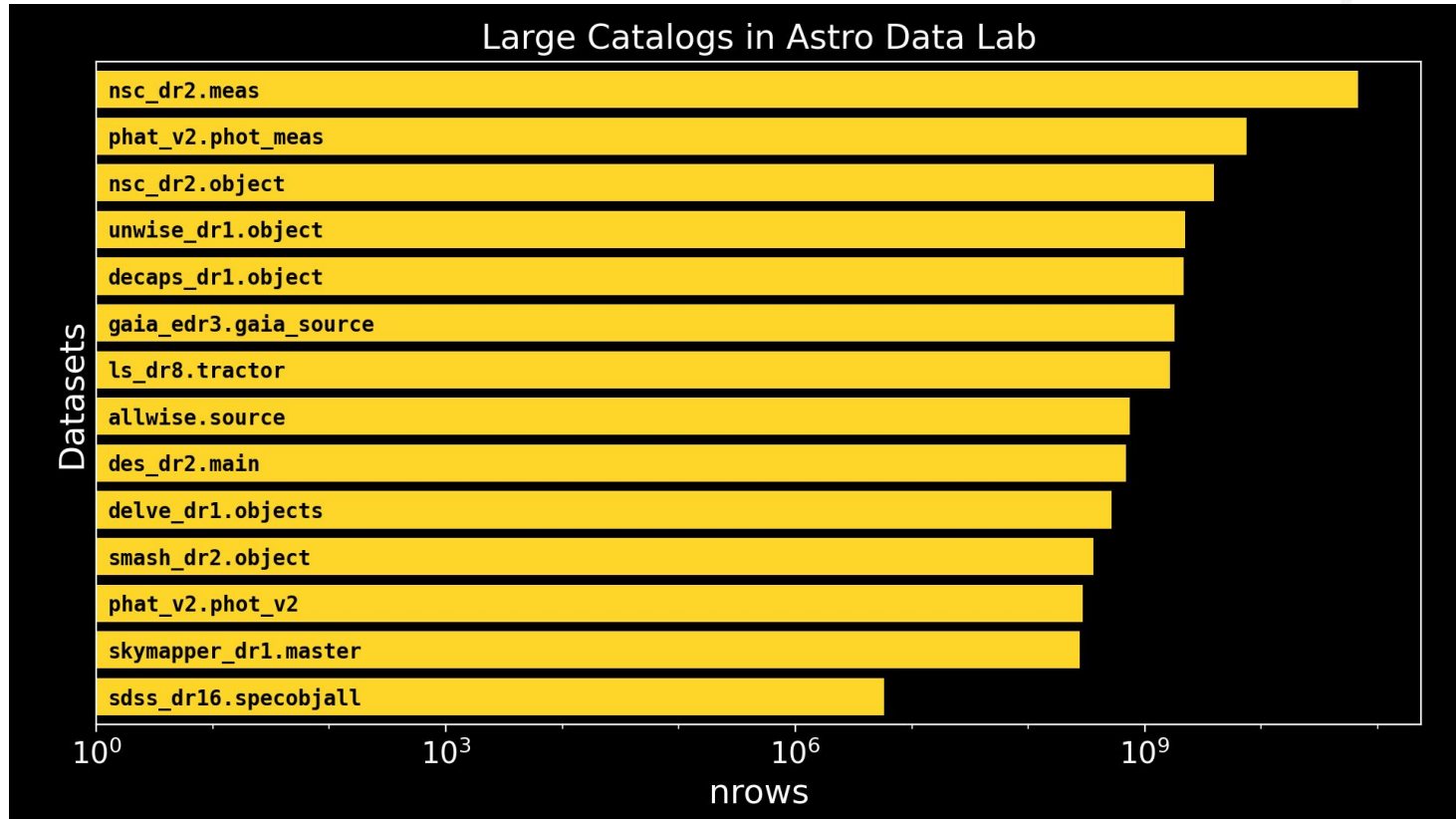


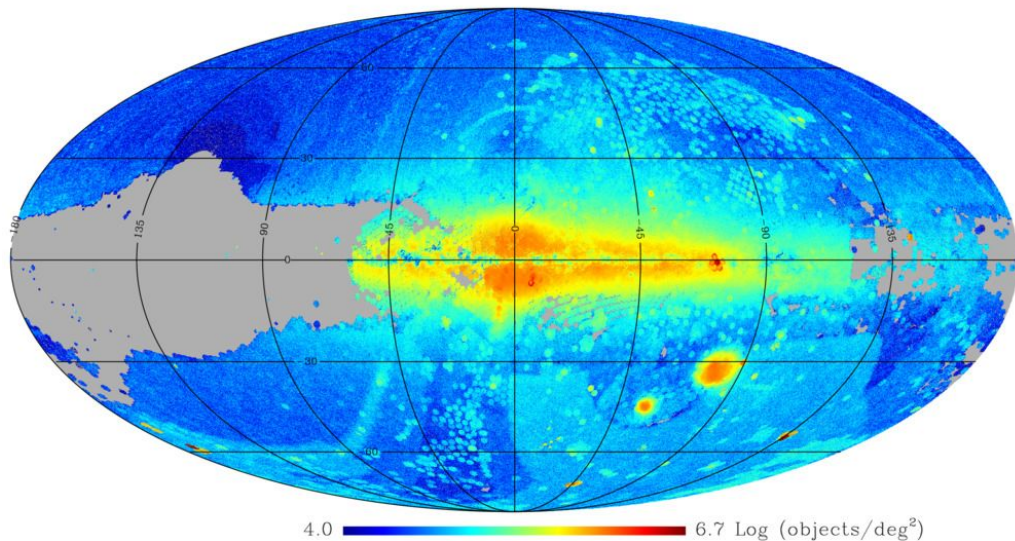
High-level goals



- Enable easy exploration of very large data holdings
→ catalogs, pixels, spectra, survey file collections...
- Connect the various data products, joint analysis
→ e.g. find interesting objects in catalogs, now find good images of them
- Enable remote analysis
→ Bring your code & algorithms & your data to the Big Data;
execute code on our servers; analyze; visualize; publish
- Enable easy user collaboration
→ sharing of query results, data sets, notebooks, group databases & storage

Some large tables at Data Lab



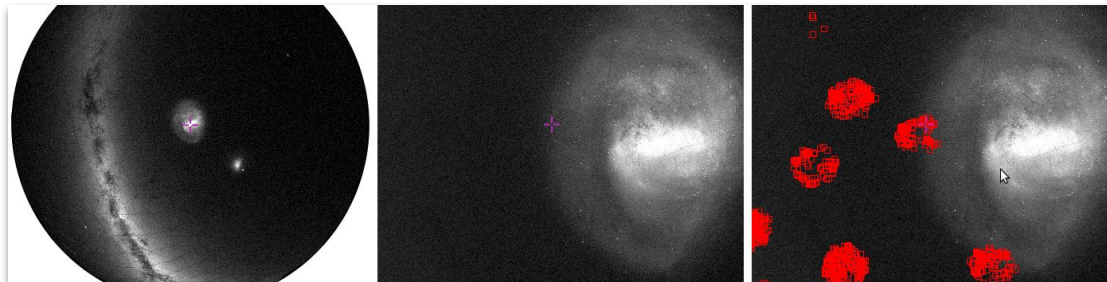


Nidever+2021 (AJ)

- 35,273 sq deg
 - u,g,r,i,z,U,VR bands
 - 412,116 exposures
 - 3.9 billion objects
 - 69 billion measurements
 - 100s of epochs in some regions
 - 21-23.6 mag median depth
 - 0.99-1.35 arcsec median seeing
 - Photometric calibration accuracy 1-2%
 - Astrometric calibration accuracy 2 mas
-
- DR3 with PSF photometry mid-2022

(Visual) data exploration

Web survey viewer
(based on Aladin Lite)

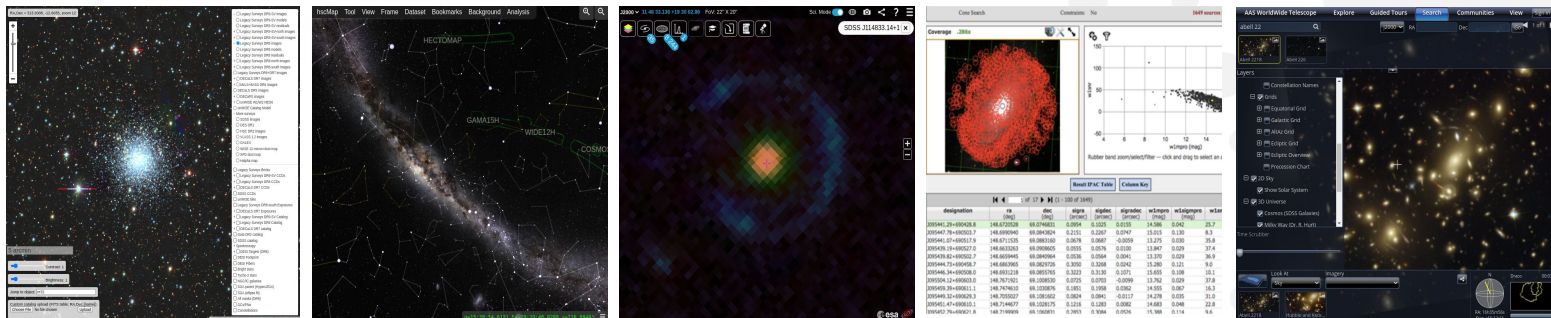


Survey footprint navigation Pan & Zoom
(here on the LMC)

Catalog overlays
(here SMASH fields)

Also looking into other viewers, e.g.

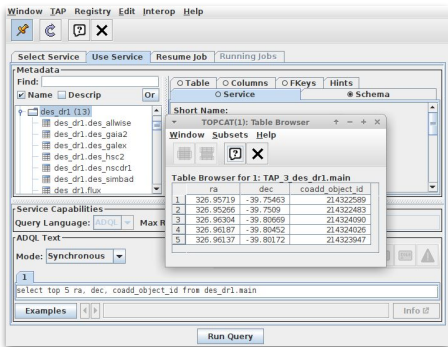
- Legacy Sky Viewer
- hscMap
- ESA Sky
- Firefly
- WWT
- ...



Getting to catalog data (1)

- SQL-like queries via TAP → PostgreSQL and ADQL
- Both sync and async queries → Submit & wait / Submit & check later
- Can query both DL catalog holdings and user's MyDB
- Clients:

TAP-aware (e.g. TOPCAT)



queryClient.py (notebooks, scripts)

```
1 query = 'select ra, dec, coadd_object_id from des_dr1.main limit 5'
2 print(qc.query(query))
```

```
ra,dec,coadd_object_id
326.957189, -39.754627, 214322589
326.952661, -39.750899, 214322483
326.963039, -39.806693, 214324090
326.96187, -39.804521, 214324026
326.961371, -39.801715, 214323947
```

datalab CLI (on local computer)

```
$datalab query sql="select ra, dec, coadd_object_id from des_dr1.main limit 5"
ra,dec,coadd_object_id
326.957189, -39.754627, 214322589
326.952661, -39.750899, 214322483
326.963039, -39.806693, 214324090
326.96187, -39.804521, 214324026
326.961371, -39.801715, 214323947
```

Query from DL website

Results 1-5 of 5 (5 before filtering) Show 10 results per page

Text boxes under each column define filters to select rows matching the condition (e.g. <26.59:00) Apply Filter Clear Filter

Select All Rows Unselect All Rows Show Row 2 Values

ra	dec	coadd_object_id
Number	Number	Number
<input type="checkbox"/> 326.95718899999997	-39.754626999999999	214322589
<input type="checkbox"/> 326.95266099999998	-39.750889999999997	214322483
<input type="checkbox"/> 326.96303899999998	-39.806692999999996	214324090
<input type="checkbox"/> 326.96186999999998	-39.804521000000001	214324026
<input type="checkbox"/> 326.96137099999999	-39.801715000000002	214323947

Getting to catalog data (2)

```
qc.query('select * from gaia_dr2.gaia_source',  
out='?') # ? = mydb://<tablename> | vos://<filename>
```

- Query results can be either:
 - Streamed back to client → *Convert yourself, according to your needs*
 - Loaded as table straight to user's MyDB → *Great for subsequent x-matching*
 - Saved as file to user's VOspace → *Great for sharing as a file with others*
- Output can be streamed back in various formats, e.g.
 - CSV stream
 - Pandas data frame
 - VOTable, AstropyTable
 - Numpy array, record array, ...

```
result = qc.query('select * from gaia_dr2.gaia_source',\  
format='?') # ? = csv|ascii|array|structarray|table|pandas|fits|votable
```

Getting to catalog data (3)

- Query performance is key
- Optimized PostgreSQL (config), optimized tables (index cols, dtype stacking)
- Fast H/W is paramount (throughput)
- In summer 2019 we switched to SSD-based systems

Relative DB H/W Performance HDD / SSD system																												
1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20	21	22	23	24	25	26	27	28	29
Full-table queries HDD system																												
Full-table queries SSD system																												
Positional cross-matches HDD																												
Positional cross-matches SSD																												
Number of parallel big queries HDD																												
Number of parallel big queries SSD																												

We are now purely CPU-bounded up to ~15 sustained large queries running in parallel; a good place to be!

- Expanding SSD storage again now (to ~150 TB, RAID-6)

Connecting catalogs with catalogs: Cross-matching

Python API

(e.g. in Jupyter notebooks)

Positional cross-matching web tool

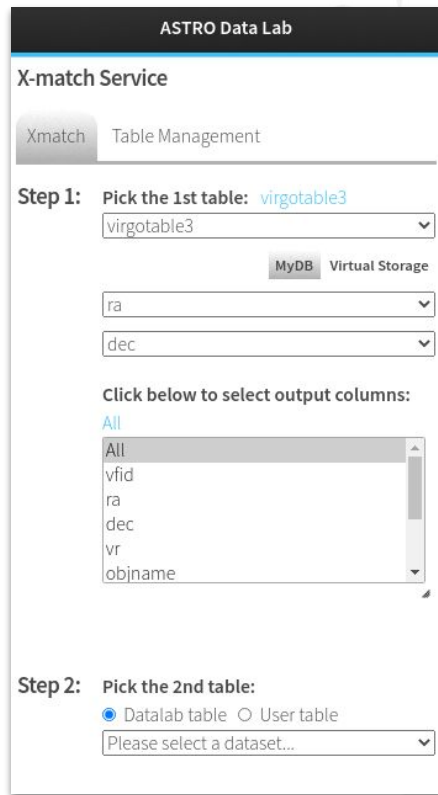
(uses same API)

On the backend: (Quad Tree Cube, Q3C)

Very fast! (Koposov & Bartunov 2006)

Tens of millions of rows in user table

are no problem



The screenshot shows the 'ASTRO Data Lab' interface for the 'X-match Service'. It has two tabs: 'Xmatch' (selected) and 'Table Management'. Under 'Step 1: Pick the 1st table:', the dropdown menu is set to 'virgotable3'. Below this, there are two dropdown menus for 'ra' and 'dec'. There are also buttons for 'MyDB' and 'Virtual Storage'. A section titled 'Click below to select output columns:' contains a list box with 'All' selected, and other options: 'vfid', 'ra', 'dec', 'vr', and 'obiname'. Under 'Step 2: Pick the 2nd table:', there are radio buttons for 'Datalab table' (selected) and 'User table', followed by a dropdown menu with the text 'Please select a dataset...'.

Connecting catalogs with catalogs: Cross-matching / Pre-computed x-match tables

- Originally: on a what-makes-sense-per-survey basis
- Now working toward an (almost) automated mechanism:
 - For each new data set, x-match against all reference sets
 - Astrometry*: latest Gaia `gaia_edr3.gaia_source`, `gaia_dr2.gaia_source`
 - Photometry*: latest NSC `nsc_dr2.object`, `unwise_dr1.object`
 - Spectroscopy*: latest `sdss_drN.specobjall` (currently N=16)
 - Default matching radius (1.5"), single nearest neighbor, no empty rows
 - Match table has few columns: `id1`, `ra1`, `dec1`, `id2`, `ra2`, `dec2`, `angular distance`
 - Re-compute when reference data sets are updated

Connecting catalogs with pixels: SIA service and image cutout

Simple Image Access:

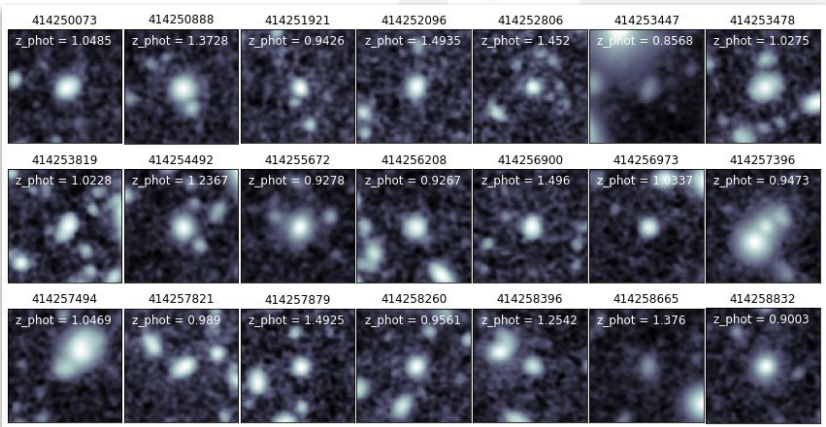
- Query metadata DB about images that contain RA/Dec position
- Constraints are possible (i.e. exposure time, product type, ...)
- **access_url** field has link to FITS file
- **fov** argument used to compute a (usually) square cutout
- We query own image holdings & those at the NOIRLab *Astro Data Archive*

```
from pyvo.dal import sia
svc = sia.SIAService("https://datalab.noirlab.edu/sia/gogreen_dr1")

imgTable = svc.search((ra,dec), (fov, fov), verbosity=2).to_table().to_pandas()
imgTable
```

	assoc_id	access_url	access_format	access_estsize	dataprodtype	dataprodsubtype
0	b'gogreen_dr1'	b'http://datalab.noao.edu/svc/cutout?col=gogre...	b'image/fits'	111924	b''	b''
1	b'gogreen_dr1'	b'http://datalab.noao.edu/svc/cutout?col=gogre...	b'image/fits'	111924	b''	b''

2 rows x 61 columns



User data storage services

Co-location of user data and DL holdings

User file storage: VOSpace

- 1 TB / user (soft quota)
- read/write, access via `storeClient.py` and `dataLab` CLI
- `public/` subdirectory to share files with other users
- read-only linked in user's Jupyter notebook space

User database: MyDB

- 250 GB / user (soft quota)
- read/write, access via `queryClient.py` and `dataLab` CLI
- also used for very fast positional cross-matching

Public (read-only) file services to serve heterogeneous survey file collections, e.g.

- Arbitrary directory structure
- Weight masks, images, catalog files
- Documentation files
- “Aux” files... anything goes

Access through *storeClient.py* and *dataLab* CLI

```
print(sc.services())
```

name	svc	description
chandra	vos	ChAMPlane: Measuring the Faint X-ray Bin ...
cosmic_dawn	vos	Cosmic DAWN survey
deeperange	vos	Deeperange Survey
deep_ecliptic	vos	Deep Ecliptic Survey
dls	vos	Deep Lens Survey
flamex	vos	FLAMINGOS Extragalactic Survey
fls	vos	First Look Survey
fsvs	vos	Faint Sky Variability Survey
ir_bootes	vos	Infrared Bootes Imaging Survey
lgs	vos	Local Group Survey
gogreen_dr1	vos	GOGREEN DR1 Survey
lmc	vos	SuperMACHO Survey
ls_dr1	vos	DECam Legacy Survey DR1
ls_dr2	vos	DECam Legacy Survey DR2
ls_dr3	vos	DECam Legacy Survey DR3
ls_dr4	vos	DECam Legacy Survey DR4
ls_dr5	vos	DECam Legacy Survey DR5
ls_dr6	vos	DECam Legacy Survey DR6
ls_dr7	vos	DECam Legacy Survey DR7
ls_dr8	vos	DECam Legacy Survey DR8
m31_newfirm	vos	M31 NEWFIRM Survey
ndwfs	vos	NOAO Deep-Wide Survey
nfp	vos	NOAO Fundamental Plane Survey
nmbs	vos	NEWFIRM Medium Band Survey
nmbs_2	vos	NEWFIRM Medium Band Survey II
nsc	vos	NOAO Source Catalog
sdss_dr8	vos	SDSS DR8
sdss_dr9	vos	SDSS DR9
sdss_dr10	vos	SDSS DR10
sdss_dr11	vos	SDSS DR11
sdss_dr12	vos	SDSS DR12
sdss_dr13	vos	SDSS DR13
sdss_dr14	vos	SDSS DR14
sdss_dr15	vos	SDSS DR15
sdss_dr16	vos	SDSS DR16
singg	vos	Survey for Ionization in Neutral-Gas Gal ...
smash_dr1	vos	SMASH DR1
smash_dr2	vos	SMASH DR2
sze	vos	SZE+Optical Studies of the Cosmic Accele ...
w_project	vos	The w Project
zbootes	vos	z-band Photometry of the NOAO Deep-Wide ...

Next: massively-multiplexed spectro

Next frontier are data products from massively-multiplexed spectroscopic surveys:

- Past and now: *SDSS*, including *SDSS-V*
- Now and soon: *DESI*
- Future: *MSE*, *FOBOS*, ...

Issues:

- Traditional spectra access via FITS files is painfully slow (>1s)
- Overhead in file I/O.
- Not feasible for many spectra.

⇒ **Develop fast spectro service.**

⇒ How do we access millions of spectra quickly?



DESI Dark Energy Spectroscopic Instrument



U.S. DEPARTMENT OF
ENERGY

Office of
Science



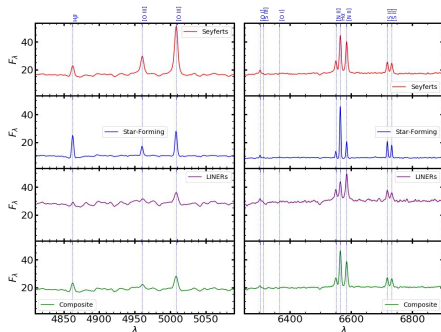
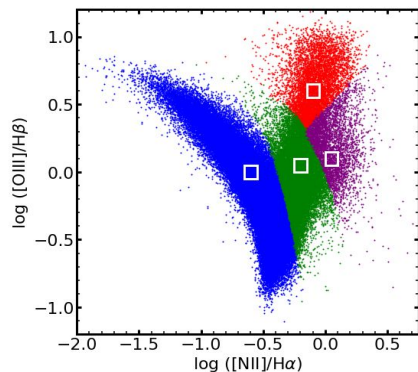
- ★ 14,000 square degrees
- ★ 40 million spectra of galaxies and quasars!
- ★ 10 million spectra of stars
- ★ Commissioning/early SV data in 2019-20
- ★ 5-year survey to start soon

Object Class	Number of Spectra	Redshift Range
bright galaxies, $r < 19.5$	10 million	$0 < z < 0.4$
luminous red galaxies (LRGs)	4.2 million	$0.4 < z < 1.0$
emission line galaxies (ELGs)	18 million	$0.6 < z < 1.6$
quasars (QSOS)	2.4 million	$0.5 < z < 3.5$
Milky Way stars	10 million	---



Mayall 4m (Kitt Peak, AZ)

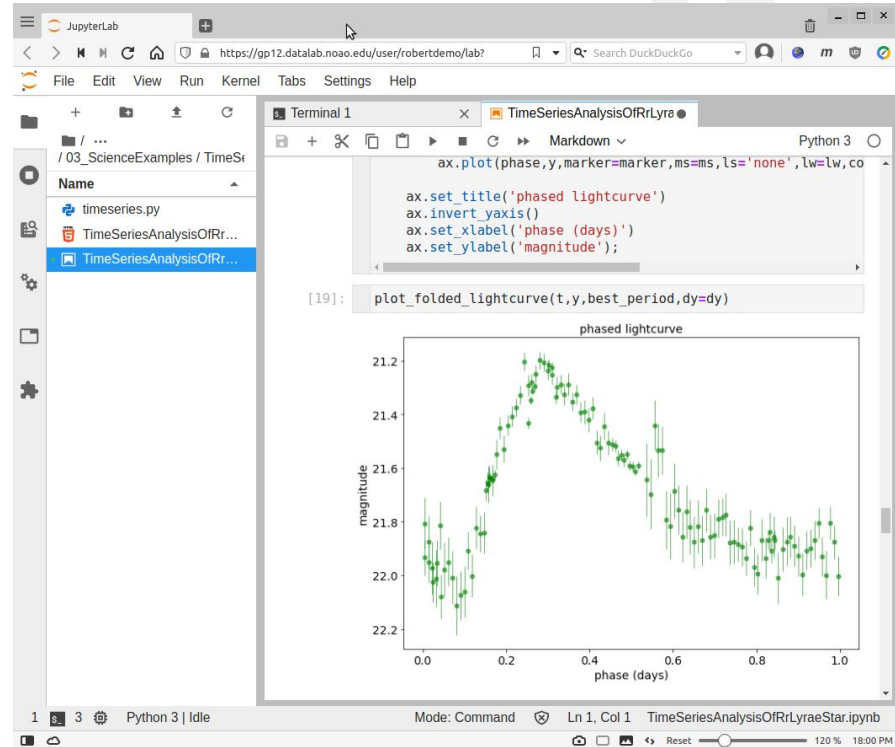
Stacking SDSS Spectra of Galaxies Selected from the BPT Diagram



- With Data Lab SDSS file service, the notebook executed in **7.5min using 4x100 spectra**
- Using the current DL Spectral Service demonstrator, notebook completes in 25 sec (**8 sec for spectrum access**)
- Currently developing production-level system that can **scale to DESI**

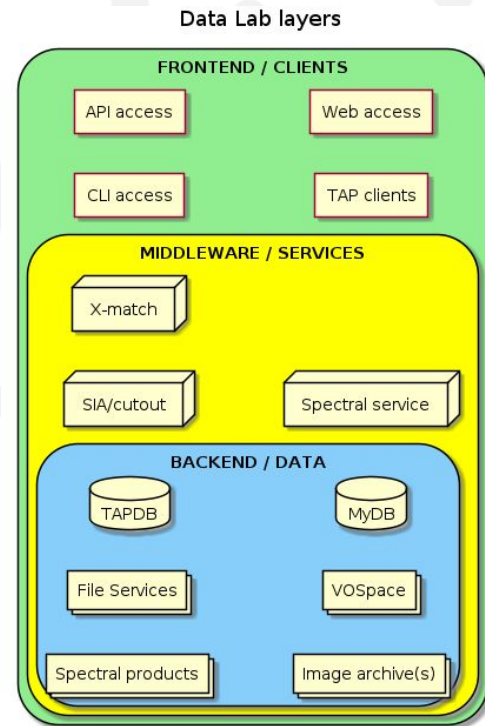
Bring your analysis to the big data

- Remote computing, co-location with data
- No installation required (just a browser)
- Jupyter notebook server
- DL-curated NB suite + user-contributed NBs
- Full astro S/W stack installed
- Planned: users own their containers, e.g.
 - can install S/W
 - can start from scratch any time
- Interfaces to data, to services, and to user storage (DB and VOSpace)



Many ways to connect to the data

- Support standard IVOA protocols wherever possible
e.g. TAP, SIA, ADQL, (SSA)
- Develop purposefully custom mechanisms where needed
e.g. PostgreSQL, Q3C
- Translate between layers where needed
e.g. Data Lab queryClient, storeClient
- Support various access modes
e.g. sync and async queries
- Support various (popular) access methods
e.g. Python API (notebooks, scripting), CLI, TOPCAT, Web portal services



Planned: self-managed user groups

To enable easy user collaboration and data sharing:

- User can create group → *Spontaneous collaborations*
- Can invite/remove others → *Membership management by group itself*
- Can assign roles → *E.g. admin, write+read, read-only*
- Can attach storage → *Group-owned VOSpace and MyDB instances*
- Admin/owner can dissolve group again → *Wrap it up when done*
- Projects can mint DOIs → *E.g. for a paper manuscript*

Solutions exist already out there. Adapt and/or emulate what's great, e.g.:

- Sci Server's group management + storage volume attachment
- CADC VOSpace 2.0 with DOI minting capability



Also a service: User support



Sign up for a free Data Lab account: <https://datalab.noirlab.edu/>

Get help from the DL team (**we solve every help request**)

Email: datalab@noirlab.edu

Helpdesk: <https://datalab.noirlab.edu/help>

User Manual: <https://datalab.noirlab.edu/docs/manual>

Code base: <https://github.com/noaodatalab/>

Suite of example NBs: <https://github.com/noaodatalab/notebooks-latest/>

Ping us on Twitter: @DataLabAstro

If needed: install Data Lab clients and CLI on local computer:

```
pip install noaodatalab ← Still 'noao' in the name
```



The three things to take away



- A Science Platform combines big data, co-located (remote) compute, data discovery, easy data access, analysis, visualization, and collaborative working.
- As part of the larger NOIRLab data ecosystem, the *Data Lab Science Platform* does all these, while hosting one of the largest collections of photometric data, image datasets, and adding spectroscopic capabilities. Importantly, users can *access all data products from raw to HLSP*.
- We are a *community* Science Platform: users-first, open-source (client code and NB collection), open-access (most data sets), open-standards (supporting IVOA standards and interoperability).



Obrigado.
Thank you.

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