Mining Archives: needs for Machine Learning

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Users do what users do best

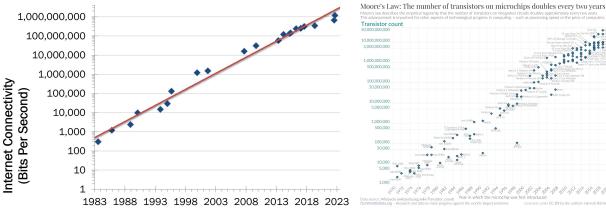


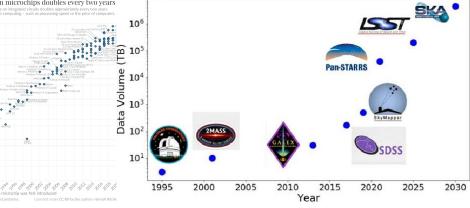
Simple Premise



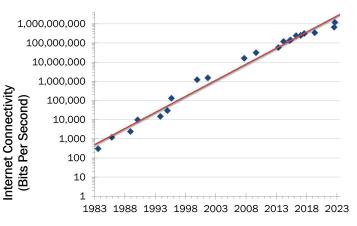
There are archives already implementing some (or even all) solutions to the problems that I will address, there is no need to jump up and say: "*I'm not guilty*"

We are all facing Data Tsunami





We are all facing Data Tsunami

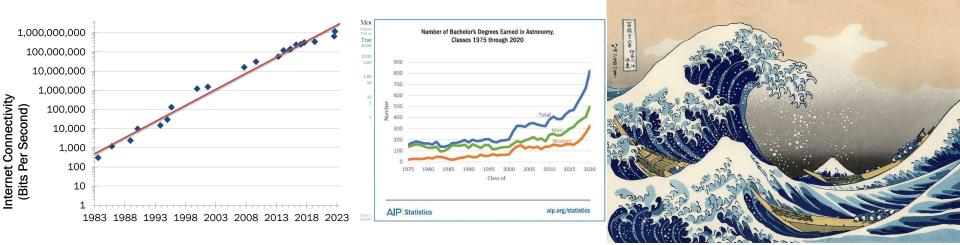


Moore's Law: The number of transistors on microchips doubles every two years Moore's underches the empirical regularity that the number of transistors on integrated crasis doubles approximately every two years. Transistor count 50000000000





Astronomers are increasing too, even though not at the same speed



More users requiring data, that are by themself increasing faster than the connection.

Transferring large datasets from astronomical archives to local machines is inefficient, consuming excessive bandwidth and storage.

One possible solution could be instead of moving data to the code, move the code to the data by providing computational resources directly within the archive infrastructure.

• This Reduced data transfer \rightarrow Minimizes bandwidth usage and speeds up analysis.

but

- Increased complexity \rightarrow Requires sophisticated infrastructure and maintenance.
- Security restrictions → Limited user access and constrained software environments.
- Reduced flexibility → Users may not have full control over the computational setup.

Some facilities already offer some computational environments to enable in-place data analysis.

This could be a viable solution but people may want to deal with more than one survey at the time, so the problem is still out there

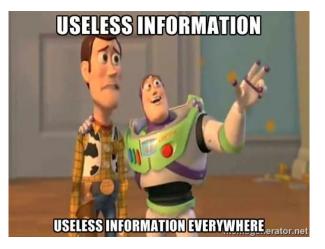


Data Tsunami Mitigation

Not all the users are interested to the same data

Just to make the easiest example:

if you want to study galaxies you don't need to retrieve data of Stars



For a typical Machine Learning experiment (actually for any kind of science) what you need is of course a **subset of data points** and a **subset of the information** available for that point.

I don't need (at least usually) to download everything is available in the archive.

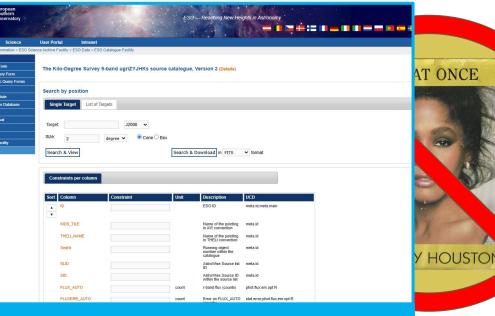
I need services that allow me to select only the data I need.



For a typical Machine Learn Readed and the second s

l don't need (at least usua is available in the archive.

I need services that allow need.



ESO portal, just as an example: I have to download every single column and I have very limited possibility to perform a complex query Even if a lot of survey are on the portal they do not "talk" one with the others

It should be obvious but:



Data quality flags are really crucial when dealing with machine learning

It is important that they are available and well documented

When comparing objects across multiple surveys, users need to cross-match them based on celestial coordinates.

This can happen on different levels:

- I have a catalogue of sources and I want the matching entries (ideal solution would be to run the cross-match on the server rather than download everything and perform the match on my side)
- I want sources with a given counterpart from a different survey, since this task could be often usual add a pre calculated ID referencing to some of the most important surveys would be useful sparing a lot of computation



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PHOTOMETRIC REDSHIFTS FOR QUASARS IN MULTI-BAND SURVEYS

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/eys, users need linates.



4 surveys involved

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Vialactea Visual Analytics Tool for Star Formation Studies of the Galactic Plane

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7 surveys involved

In order to train a deep learning model we need the images.

Not all surveys provide images and even in that case for most of them it is not straightforward to retrieve them

We are often interested in small cutout surrounding the centroid of the object but we had to download entire plates and crop them locally, leading to significant inefficiencies. Users have access to raw, full-frame data, which can be useful for context and custom processing.

Bandwidth-heavy: Downloading entire plates is costly in terms of storage and data transfer. Although it would be computationally expensive for the server this effort in terms of computing power would be balanced from the limited usage of bandwidth.

Implementing a server-side cutout service that allows users to request only the relevant portion of an image directly from the archive. Some archives like SDSS and Pan-STARRS offer this, but it is not universal and they are implemented in a very different way and they could not be exactly what a user need (going back to the initial problem).



Dataedo /cartoon

PiotreDataedo

A&A 666, A171 (2022) https://doi.org/10.1051/0004-6361/202243900 © L. Doorenbos et al. 2022 Astronomy Astrophysics

ULISSE: A tool for one-shot sky exploration and its application for detection of active galactic nuclei

Lars Doorenbos¹[©], Olena Torbaniuk^{2,3}[©], Stefano Cavuoti^{4,5}[©], Maurizio Paolillo^{2,4,5}[©], Giuseppe Longo², Massimo Brescia⁴[©], Raphael Sznitman¹, and Pablo Márquez-Neila¹[©]

~100k 73x73px 3-band thumbnails

context and custom processing.

Bandwidth-heavy: Downloading entire plates is costly in terms

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Galaxy Spectroscopy without Spectra: Galaxy Properties from Photometric Images with Conditional Diffusion Models

Lars Doorenbos^{1,7},[•], Eva Sextl^{2,7},[•], Kevin Heng², Stefano Cavuoti^{3,4}, Massimo Brescia^{3,4,5}, Olena Torbaniuk⁶, Giuseppe Longo³, Raphael Sznitman¹, and Pablo Márquez-Neila¹, Marchart, Sterner, Chella⁴, Marchart, Sterner, Fakulti für Hysik, Ludvig Maximilians Universitä Benzin Scheinestr. 1, 81679, Minchen, Gremany, sextl@usm.lmu.de ¹NRF—Astrongmical Observatory of Capodimonte, Salita Moiariello 16, 180131 Napoli, Italy ¹NRF—Astrongmical Observatory of Capodimonte, Salita Moiariello 16, 180131 Napoli, Italy ¹NRF—Science di Napoli, via Cinthia 9, 80126 Napoli, Italy ⁶Department of Physics, University Federico II, Strada Vacina Cupa Cinti, 21, 80126 Napoli, Italy ⁶Department of Physics and Astronomy "Augusto Right," University of Bologna, via Piero Gobeni 93/2, 40129 Bologna, Italy *Received 2024 June 24; revised 2024 October 16; acceptale 2024 October 25; published 2024 Docember 9*

~300k 64x64px 5-band thumbnails



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Astronomy Astrophysics

Identification of problematic epochs in astronomical time series through transfer learning*,**

Stefano Cavuoti^{1,2,***}, Demetra De Cicco^{3,1,4,***}, Lars Doorenbos⁵, Massimo Brescia^{3,1,2}, Olena Torbaniuk^{6,7}, Giuseppe Longo³, and Maurizio Paolillo^{3,1,2}

~1M 51x51px single band thumbnails

e archives like SDSS and implemented in a very reed (going back to the

I Don't Want to Miss a Thing

On the other hand I need to have control on missing data. Usually in astronomical DB a value could be missing for mainly three reason:

- A. Observed but not detected
- B. Observed but the object has been masked for a problem
- C. Not observed

It is really important to distinguish at least between A. & B. from C. since I would deal with them in a very different way.





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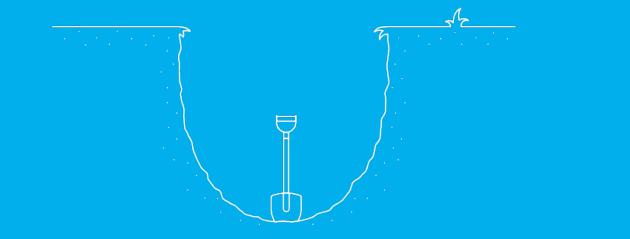
CENTRAL

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Thank you for your attention!



Possible Discussion Points

- computational environments to enable in-place data analysis
- complex query infrastructure
- importance of data quality flags
- in-place cross-match
- precomputed cross-match with surveys
- on-the-fly cutout service
- missing data information