

The logo graphic for ESCAPE features a stylized blue starburst at the top, a yellow circle in the middle, and a blue arc at the bottom, all within a white circular frame.

# ESCAPE

European Science Cluster of Astronomy &  
Particle physics ESFRI research Infrastructures

## Task 4.3 Summary and Status

Martino Romaniello – ESO

On behalf of Task 4.3

# WP4 Tasks

**Task 4.1** Integration of astronomy VO data and services into the EOSC

**Lead: Marco Molinaro (INAF)**

**Task 4.2** Implementation of FAIR principles for ESFRI data through the Virtual Observatory

**Lead: Françoise Genova (CNRS-ObAS)**

**Task 4.3** Adding value to trusted content in astronomy archives

**Co-leads: Mark Allen (CNRS-ObAS) & Martino Romaniello (ESO)**

## Task 4.3 Adding value to trusted content in astronomy archives

*Next generation functionalities for creation and publication of **high-level, value-added data products** from ESFRIs*

- Assessment and application of **new techniques** – machine learning analytics. (connection to WP3)
  - *Specific example applied to ESO archives data products: spectra, cubes, source catalogues*
- Identification of **stewardship best practices**
  - *Curation and publication of next-generation data products via ESFRI archives*
  - *Technical and human aspects*

# Major milestones

<b>D4.5</b> Release of prototype machine learning enabled archive services providing value-added content to archives (Demonstration)	Month 30 July 2021
<b>D4.8</b> Final analysis report on use of IVOA standards for FAIR ESFRI and community data and best stewardship practices for value-added data (Report) <i>(includes report of feedback on prototype services developed for D4.5)</i>	Month 40 May 2022



# Deep Learning & Data Archives

- Scope: provide archive users with novel ways to identify data
  - Beyond traditional approach of specifying query parameters
  - Revolutionary extension of recent move from instrument to data keywords (e.g. exposure time to signal-to-noise)
- Target: the different data types in the ESO Archives
  - Spectra
  - Cubes
  - Source catalogues
- Let the data speak ... on a massive scale

# Deep Learning on 1D spectra: HARPS

- Scope: Deep Learning analysis of the entire HARPS archive
  - High-resolution, high-stability spectrograph
  - Main science case: discovery and characterization of exoplanets
  - ~270,000 spectra, ~300,000 wavelengths channels each
  - 1D spectra, pipeline-reduced to high accuracy
- Different approaches
  - HITS: fully-connected autoencoder with 2 latent dimensions, down sampled spectral resolution
    - Agile architecture for speed and interactivity
  - ESO: combination of convolutional and fully-connected layers with between 4 to 8192 latent dimensions, full spectral resolution
    - Find minimal representation which preserves all the relevant information

# So now, how to turn this into user services?

- These techniques have the potential of being a game-changer, BUT:
  - The results have to be meaningful
  - Interaction with the results has to be user-friendly
- In practice:
  - What makes spectra similar?
    - What are the uses and limitations of a similarity service?
  - Can spectra be tagged with some physical properties of the celestial object?
    - If so, what are they?
    - What are the uses and limitations?
  - What is the best way to present and interact with the results?