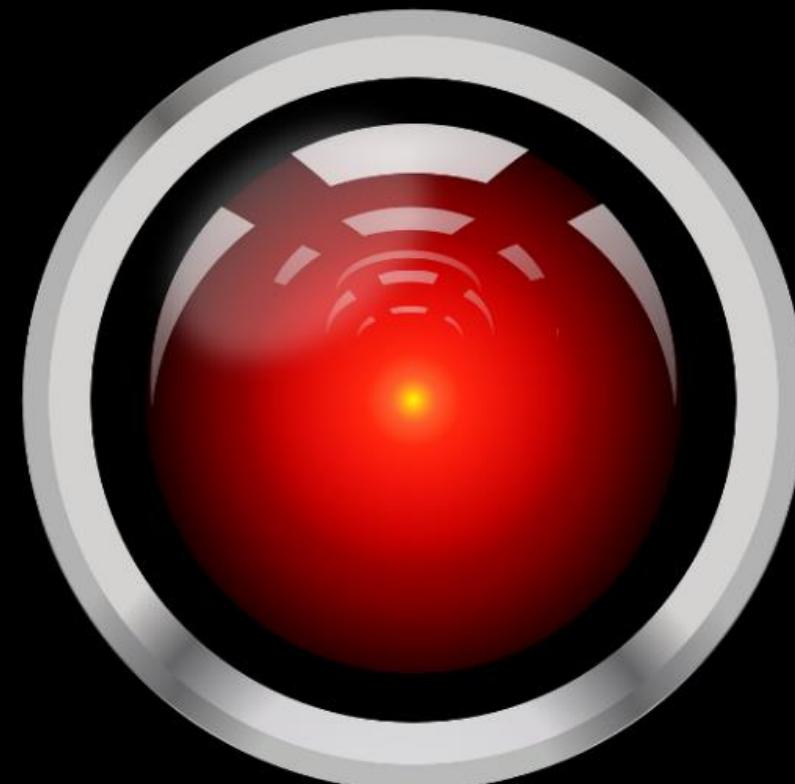
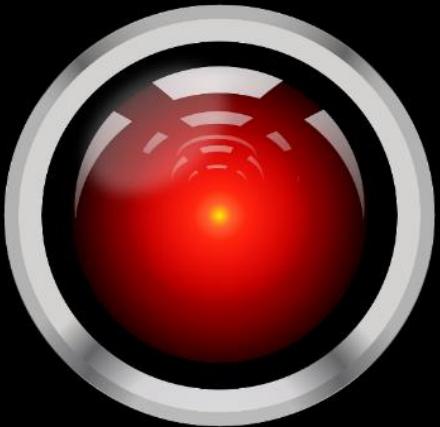


QUBRICS – finding the beacons



QUasars as
BRIght
beacons for
Cosmology
in the
Southern
hemisphere

QUBRICS-Team & References



- Konstantina Boutsia, Giorgio Calderone, Stefano Cristiani, Andrea Grazian
- Guido Cupani, Valentina D'Odorico, Fabio Fontanot, Francesco Guarneri, Matteo Porru...

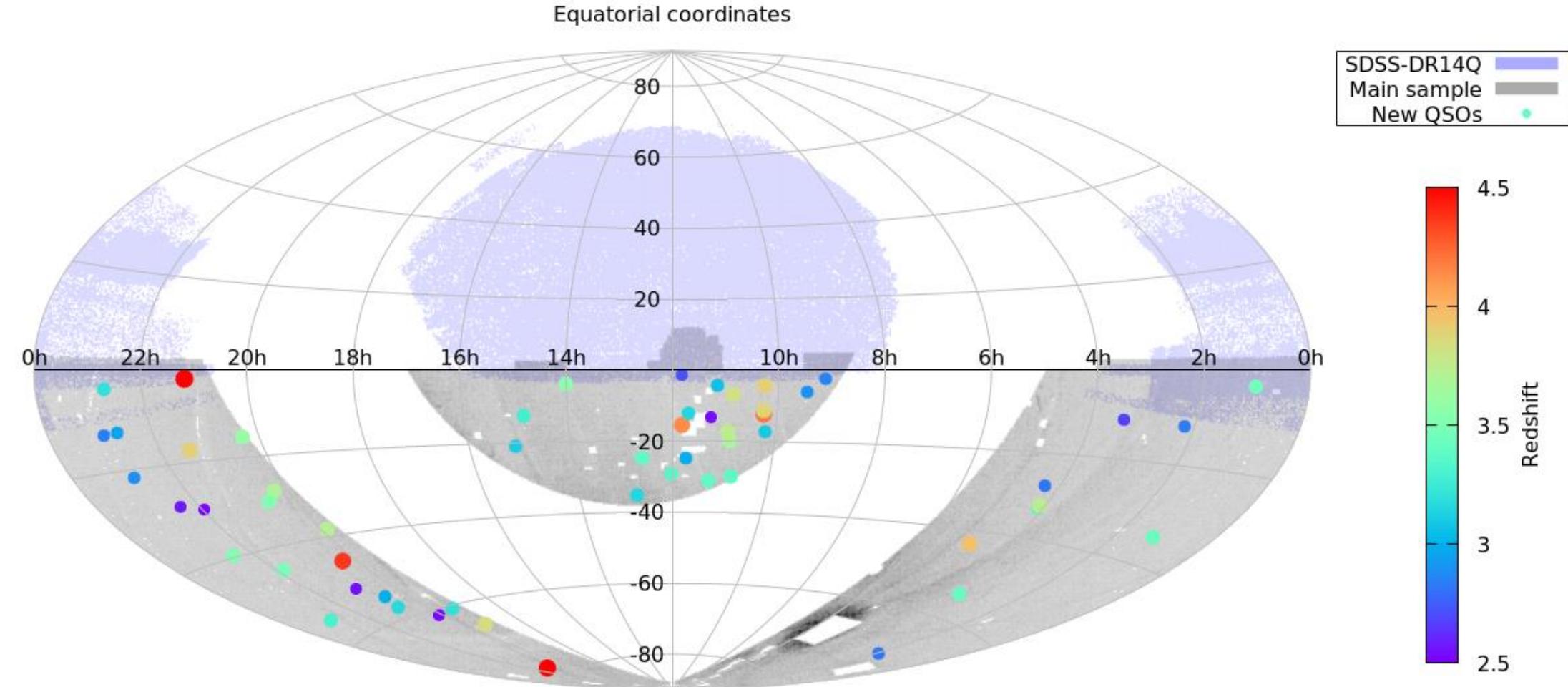
Calderone G., et al., 2019, ApJ, 887, 268...

...

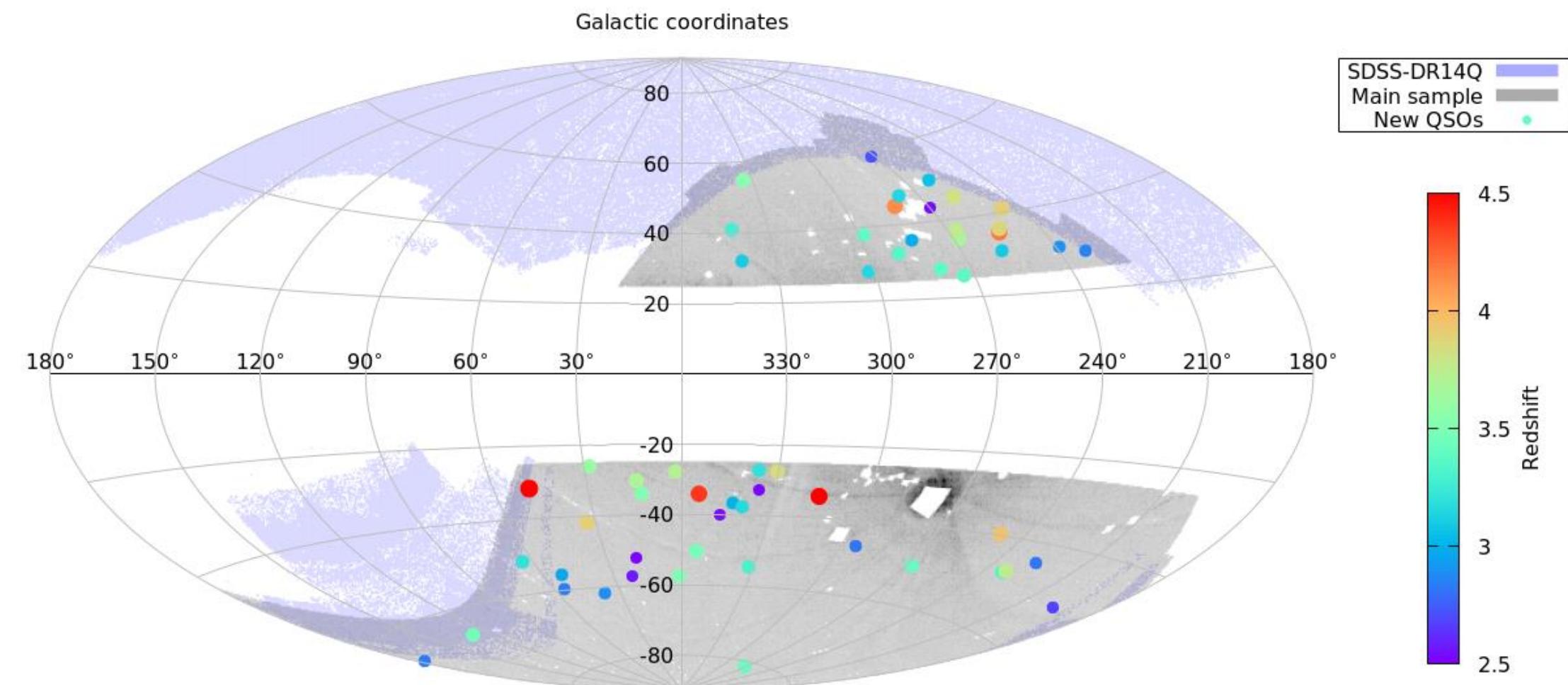
...Cristiani, S., et al., 2023, MNRAS, 522, 2019

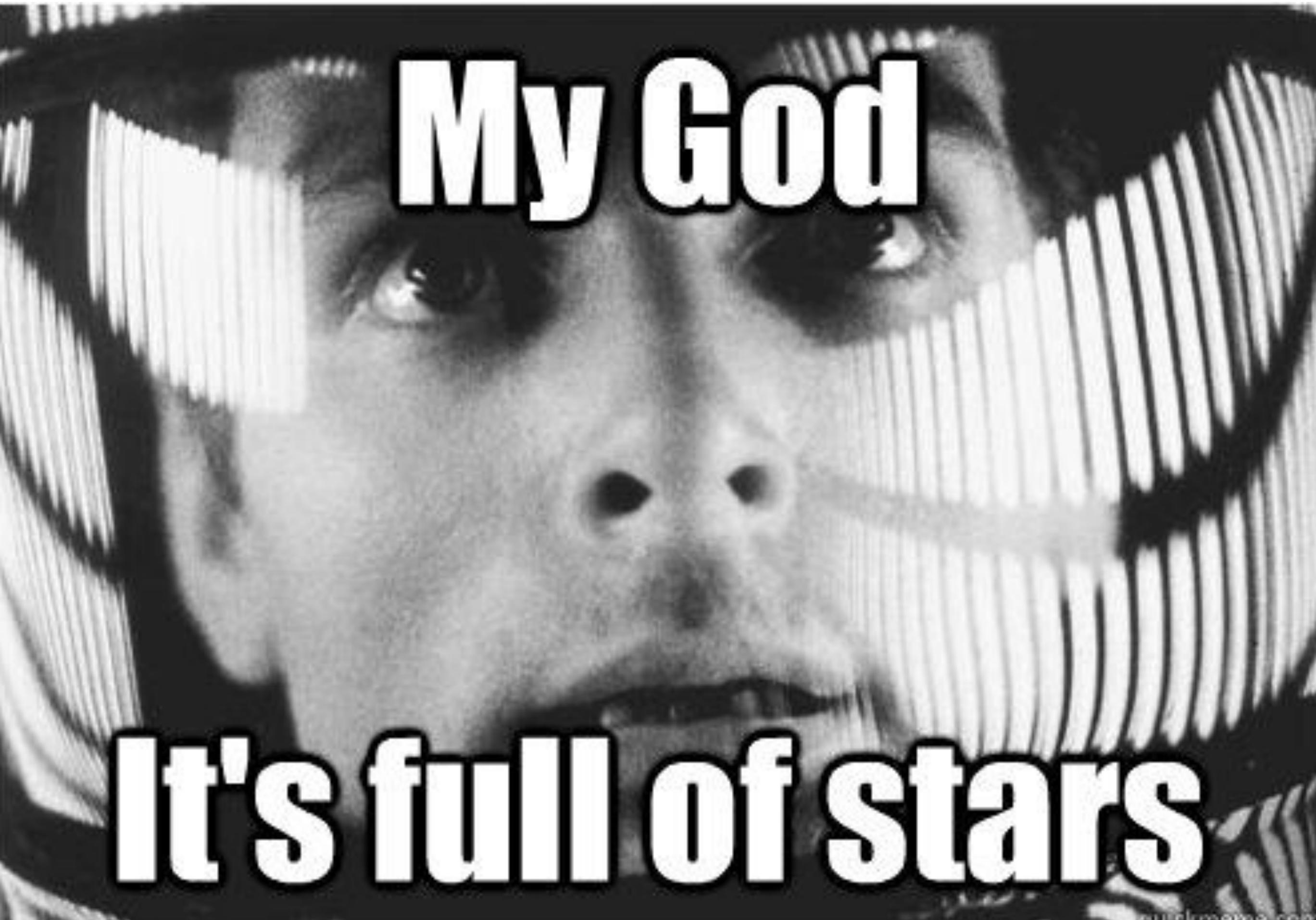
Main goal:

- Identify bright, high-redshift QSOs using data from publicly available photometric survey:
 - Gaia
 - Skymapper
 - PAN-STARRS
 - 2MASS
 - WISE
 - DES
 - eROSITA
 - Rubin Observatory
 - Euclid ...
- Two-fold problem: first identify QSOs, then remove low-redshift ($z < 2.5$) QSO/ (Classification + Interpolation)



$|b| > 25$



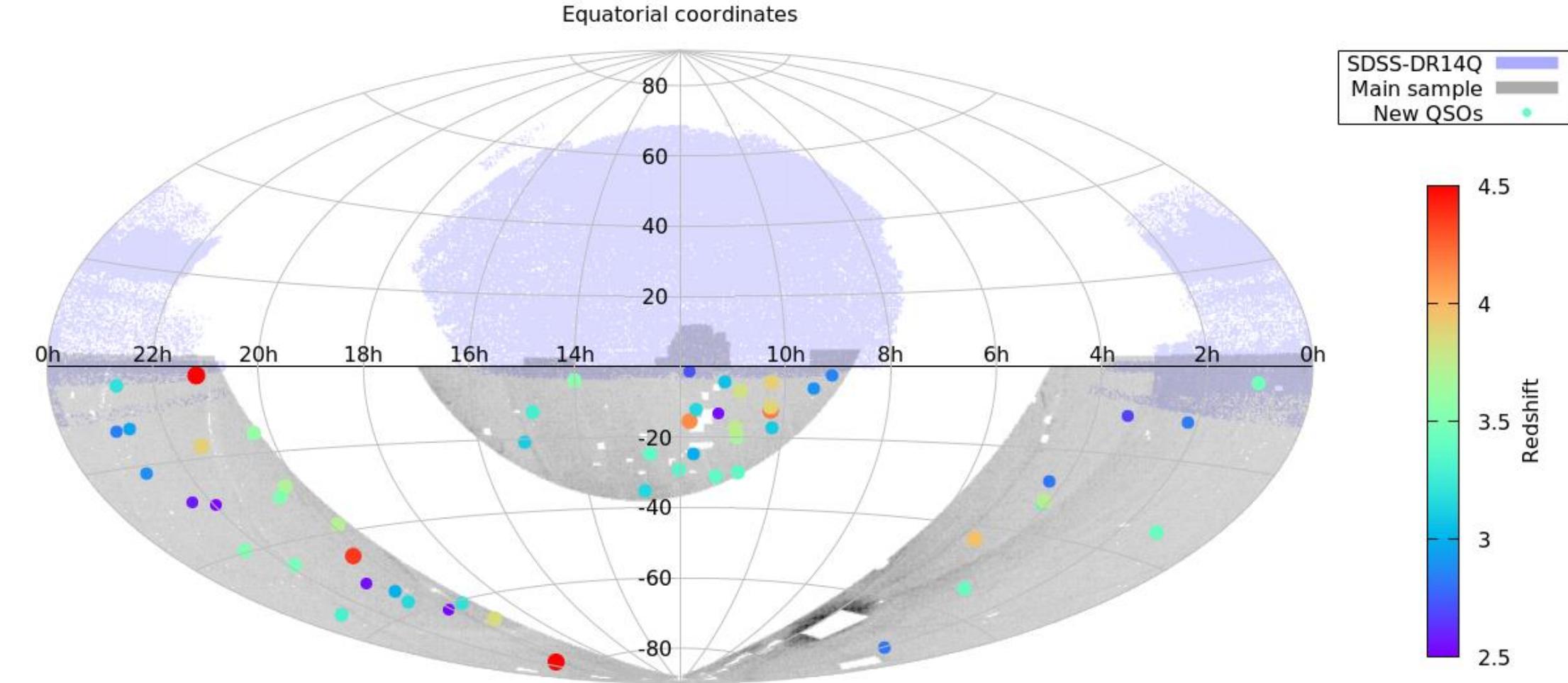
A black and white photograph of a dog's face, possibly a Shiba Inu or similar breed, looking slightly upwards and to the right. The background is a dark, textured surface that appears to be a night sky filled with stars.

My God

It's full of stars

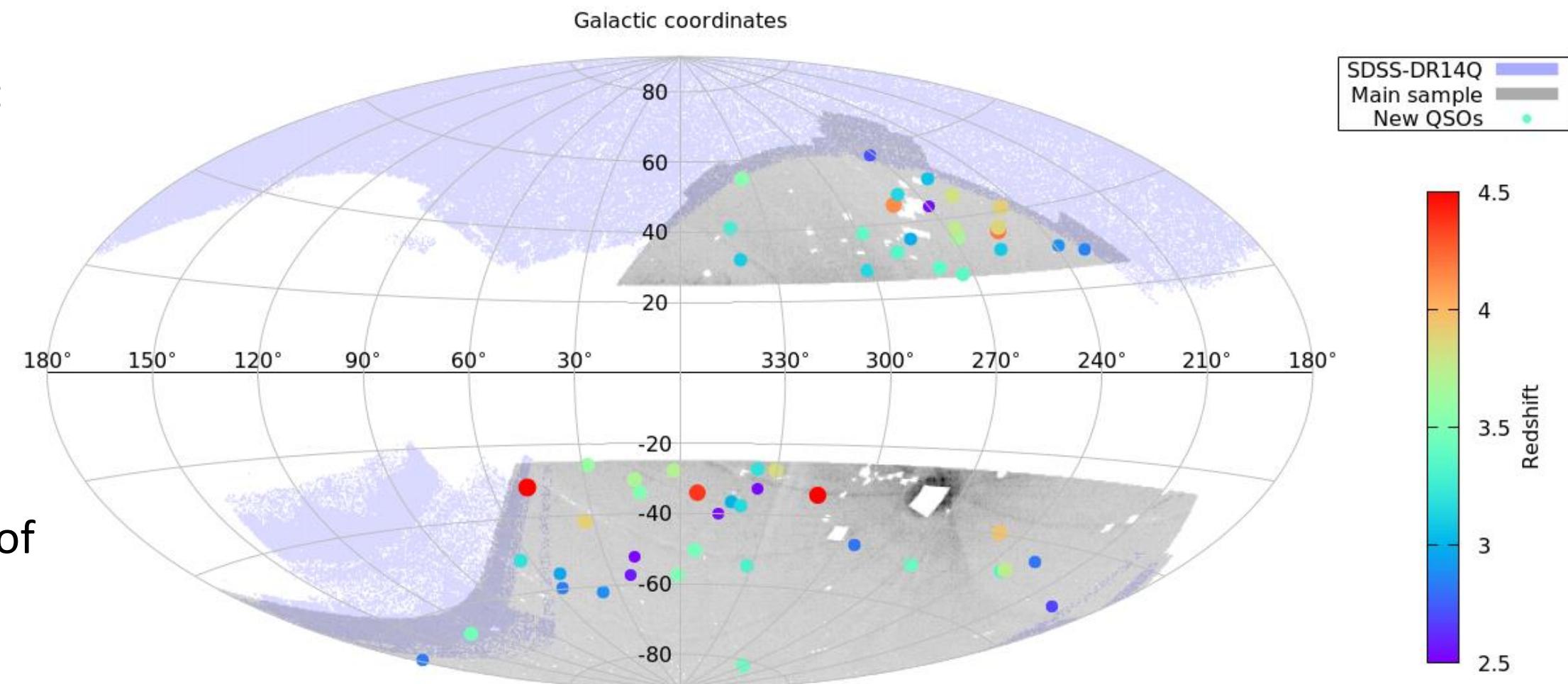
Main goal:

- Identify bright, high-redshift QSOs using data from publicly available photometric survey:
 - SkyMapper
 - Gaia
 - PAN-STARRS
 - 2MASS
 - WISE
 - DES ...
- Two-fold problem: first identify QSOs, then remove low-redshift objects



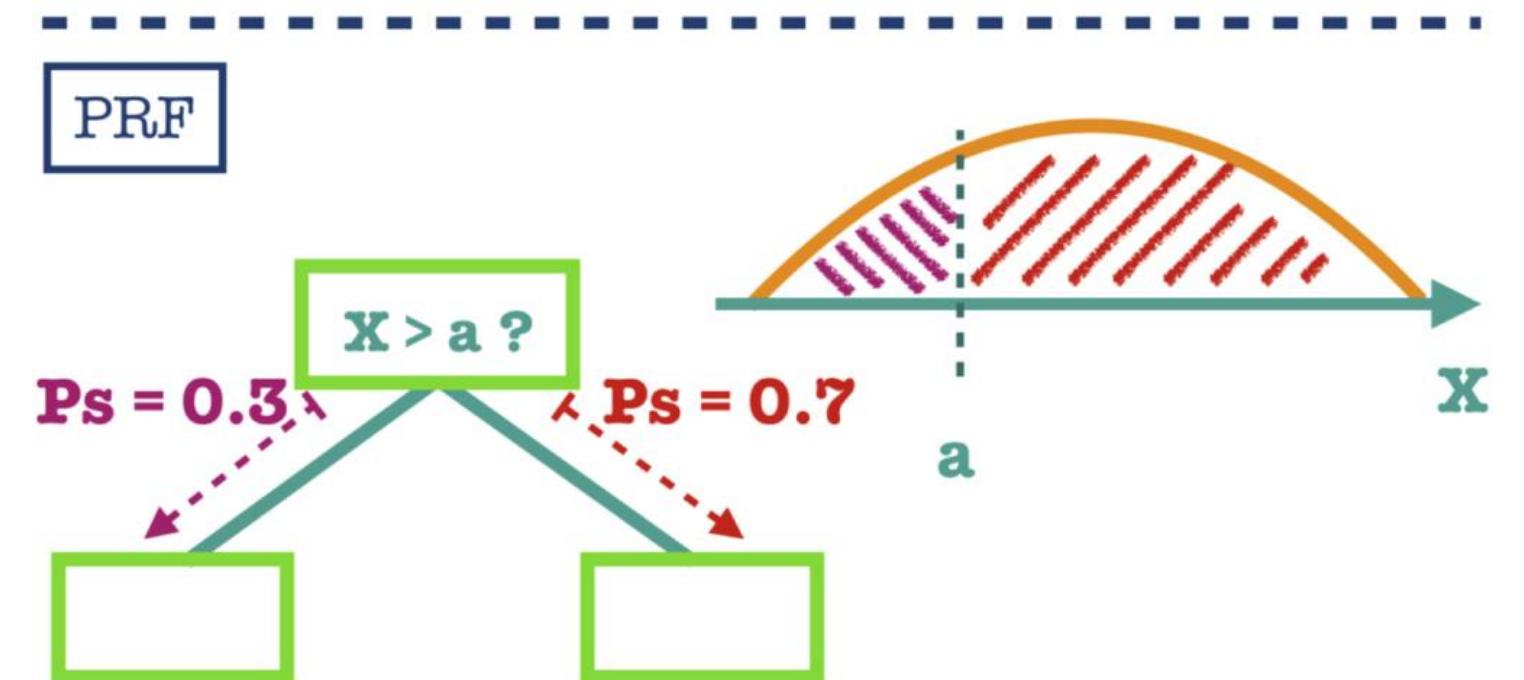
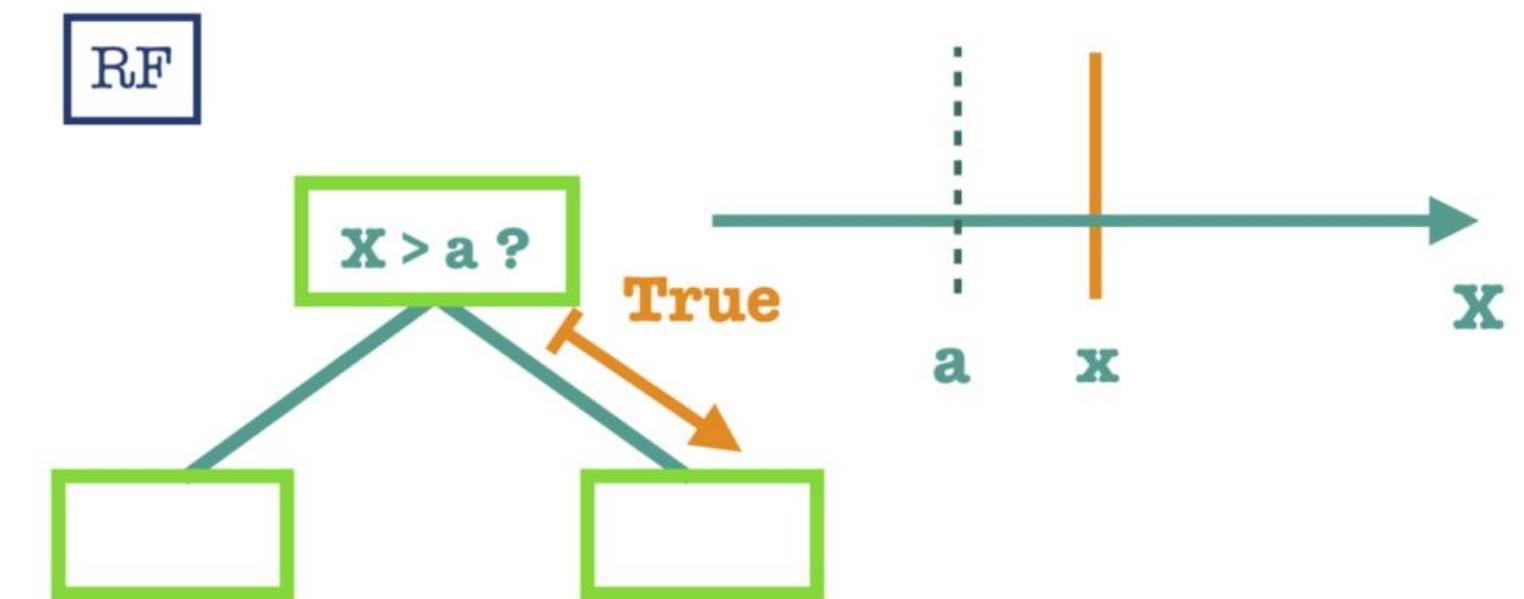
Method:

- Apply **ML techniques** on photometric datasets:
 - Canonical Correlation Analysis (**CCA**)
 - Calderone et al 2019 *ApJ* 887 268
 - Boutsia et al 2020 *ApJS* 250 26
 - Probabilistic Random Forest (**PRF**)
 - Guarneri et al 2021 *MNRAS* 506 2
 - Guarneri et al 2023 *MNRAS*
 - **XGB**: Calderone et al. 2024
- Spectroscopic follow-up to confirm the nature of high-redshift candidates



The QUBRICS survey: Probabilistic Random Forest (Reis et al. 2019)

- Generalization of the original Random Forest (RF) to account for measurement uncertainties
- In the PRF each feature is a probability distribution function: this improves performances and **takes into account errors** as variance of the distribution
- Naturally **handles missing data** (upper limits and lack of observations)!

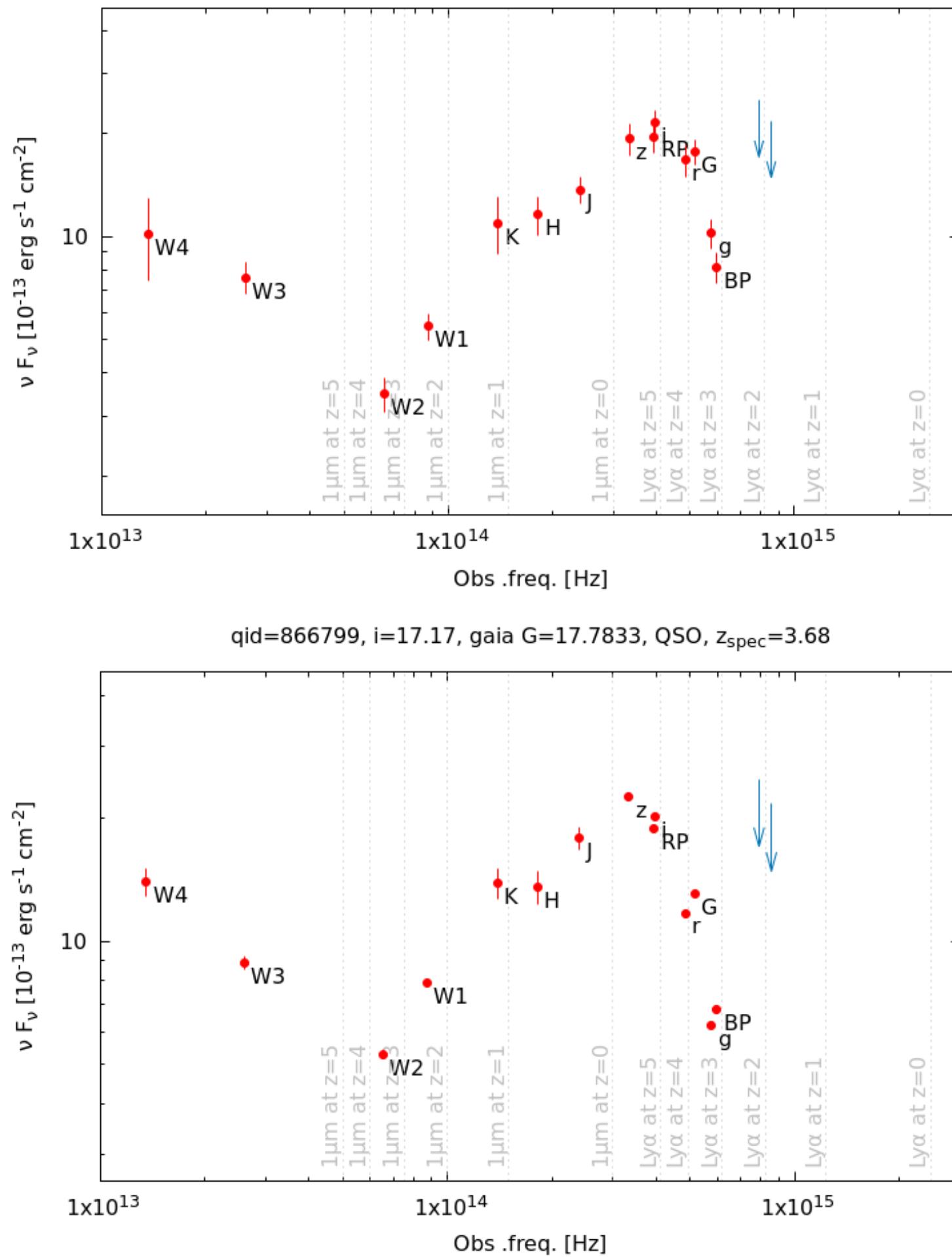


Guarneri et al 2021, MNRAS 506 2

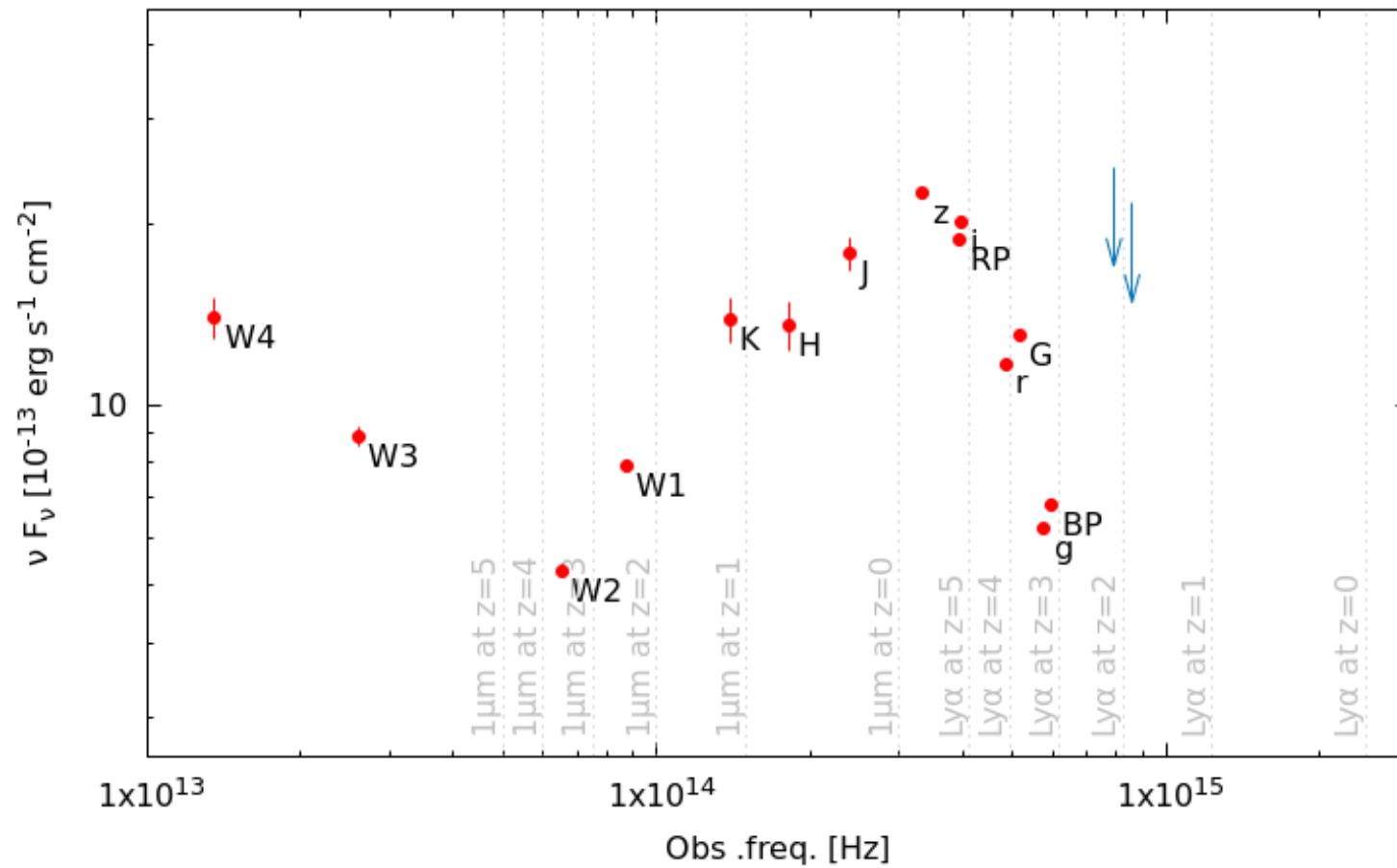
Guarneri et al 2022, MNRAS 517 2436

Reis et al. 2019 - arxiv:1811.05994

qid=210795, i=17.06, gaia G=17.456419, QSO, $z_{\text{spec}}=3.68$

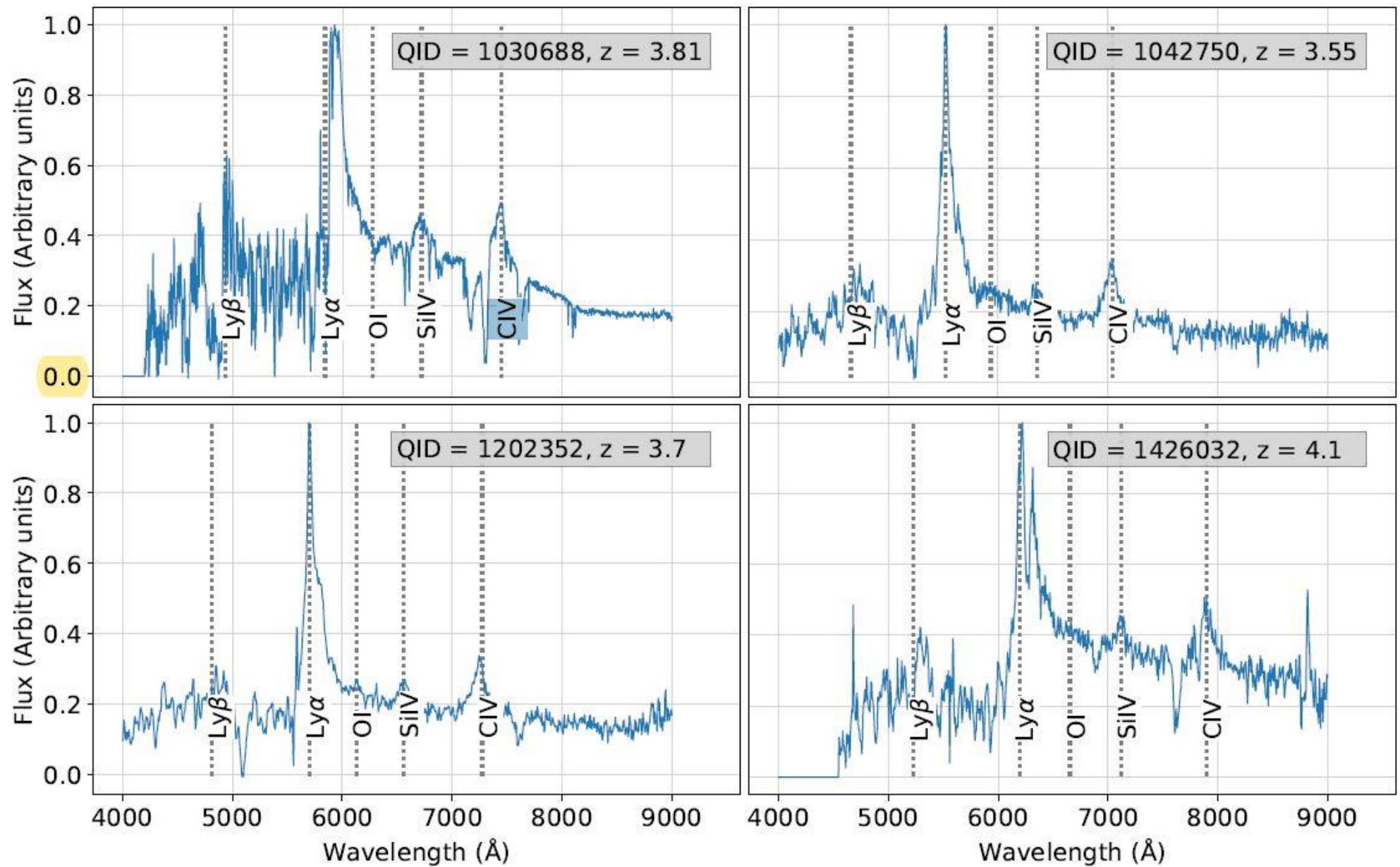


qid=866799, i=17.17, gaia G=17.7833, QSO, $z_{\text{spec}}=3.68$



Example SEDs and follow-up spectra

Guarneri et al.



Learning (many) lessons from ML

Beware of:

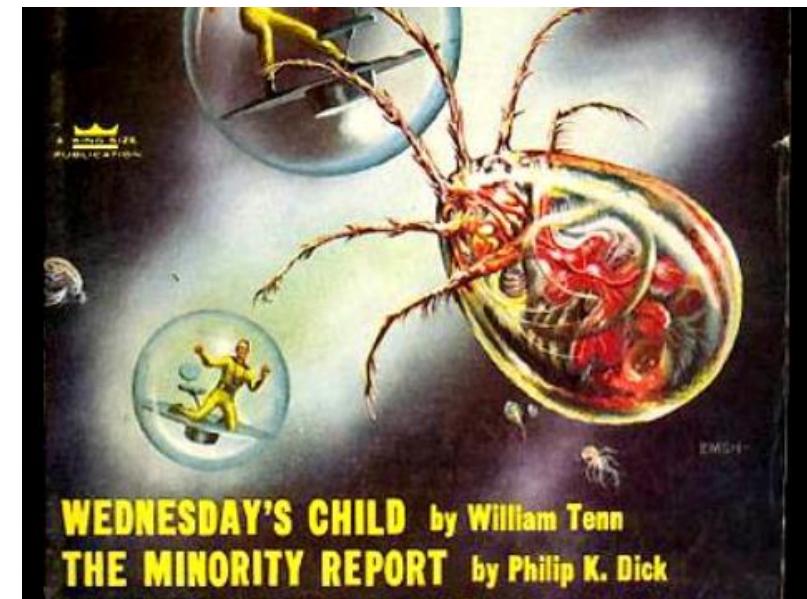
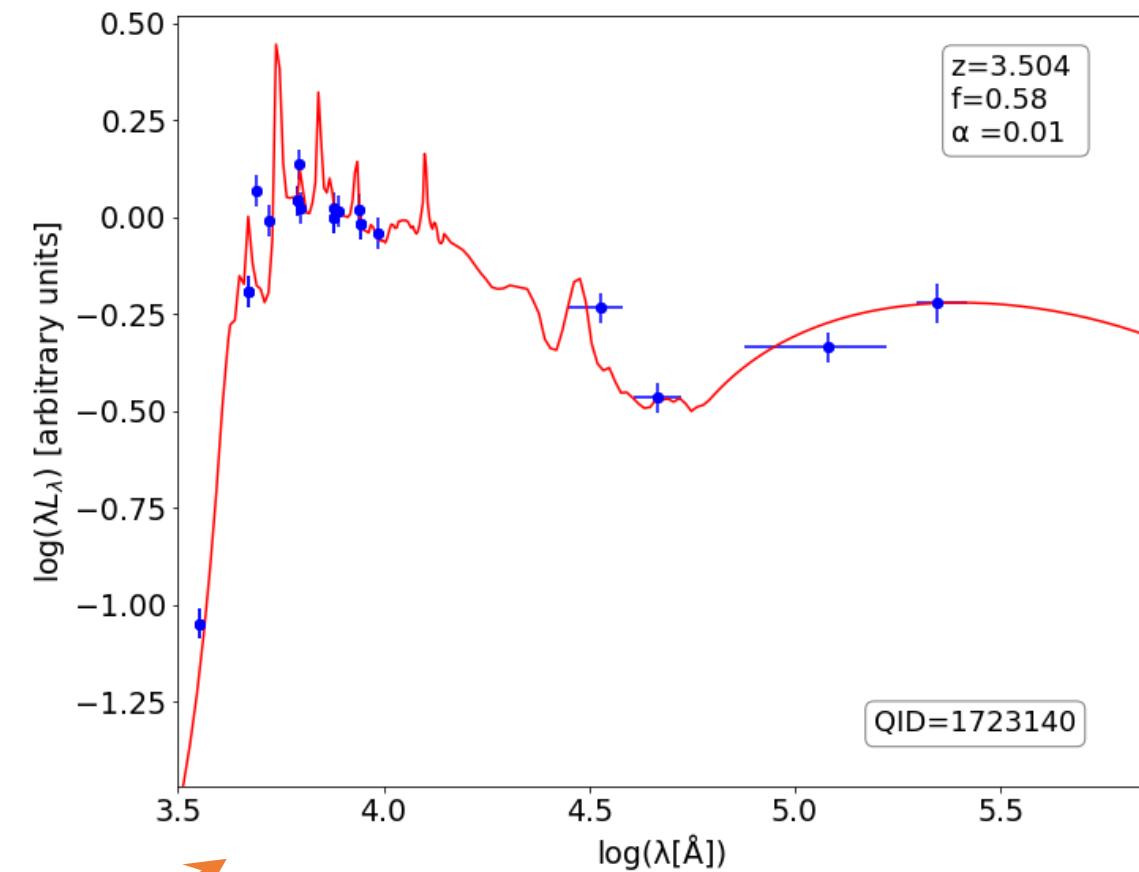
- Black box syndrome (Petch et al. 2022, e.g.)
- Overfitting (complementary methods)
- Fancy interpretation of unphysical features
- Amazing success rates and completeness

Consider that:

- Good for classification may be less good for regression
- On a well defined class, fitting a model may be fine

Need for:

- Large and balanced training sets (synthetic data)
- Proper error treatment
- Physical insight



Remember *Minority Report* (Dick, 1956)

RESEARCH ARTICLE

Open Access



Probabilistic Random Forest improves bioactivity predictions close to the classification threshold by taking into account experimental uncertainty

Lewis H. Mervin^{1*†}, Maria-Anna Trapotsi^{2†}, Avid M. Afzal³, Ian P. Barrett³, Andreas Bender² and Ola Engkvist^{4,5}

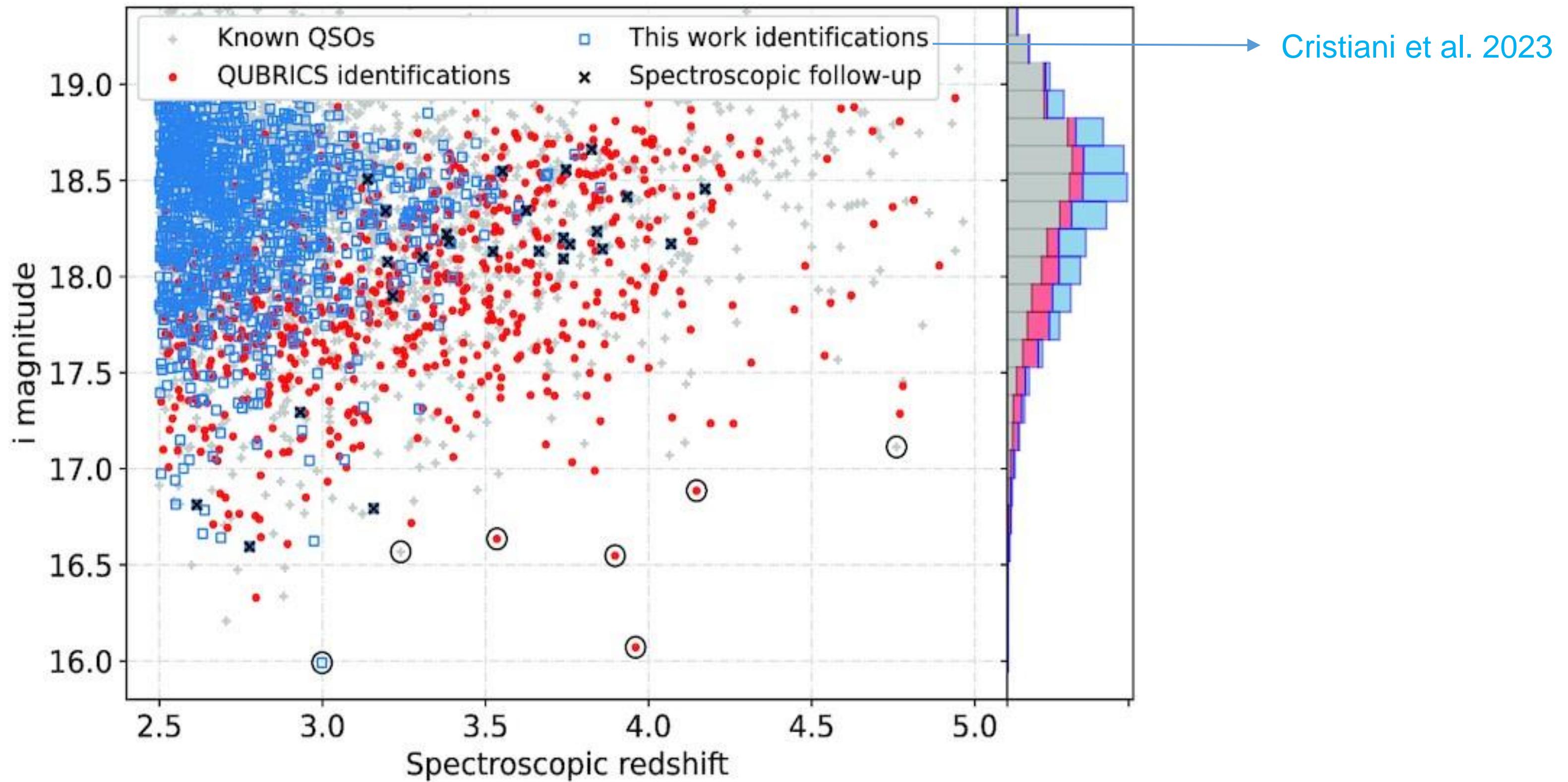
*Correspondence: lewis.mervin1@astrazeneca.com

†Lewis H. Mervin and Maria-Anna Trapotsi contributed equally to this work

¹ Molecular AI, Discovery Sciences, R&D, AstraZeneca, Cambridge, UK
Full list of author information is available at the end of the article

“It seems the nature & bias in our data shares much similarity to the ones in yours, in the search for quasars.”

Observations – Finding the cosmic beacons

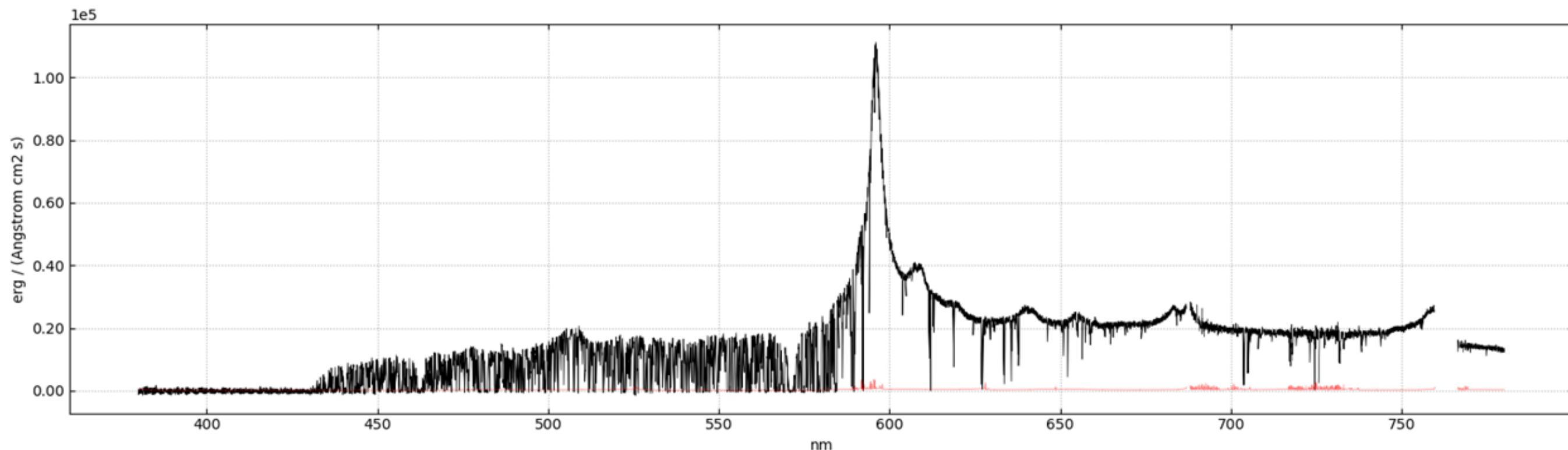


SB1

total int.: ~ 8.6 h

$\langle \text{SNR} \rangle$: ~ 60

last obs.: 15/8/23

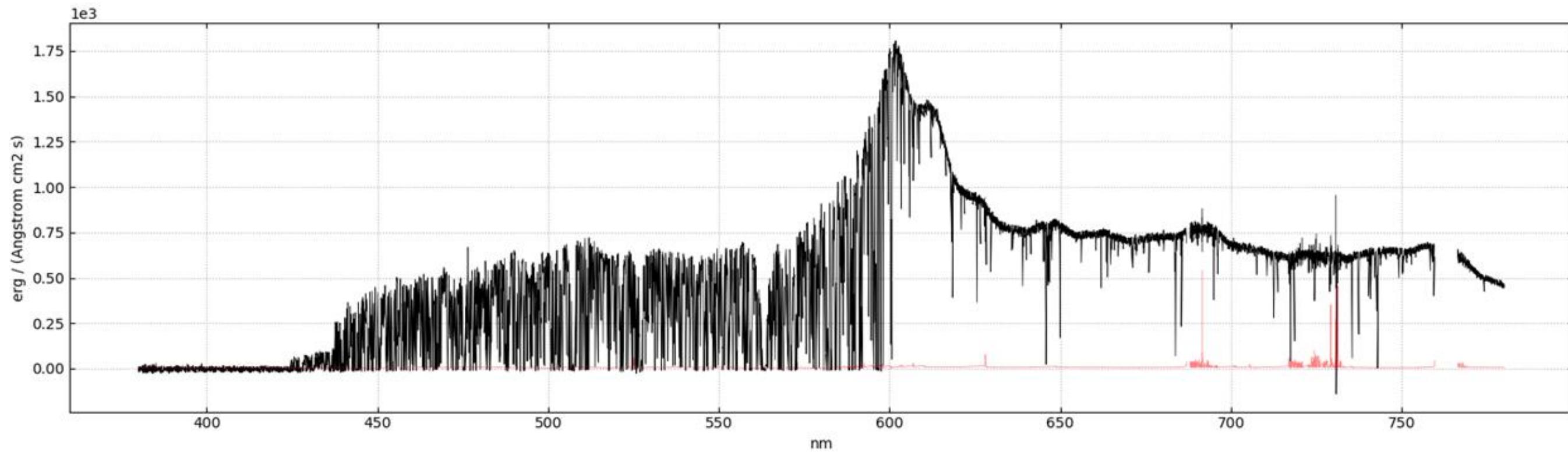


SB2

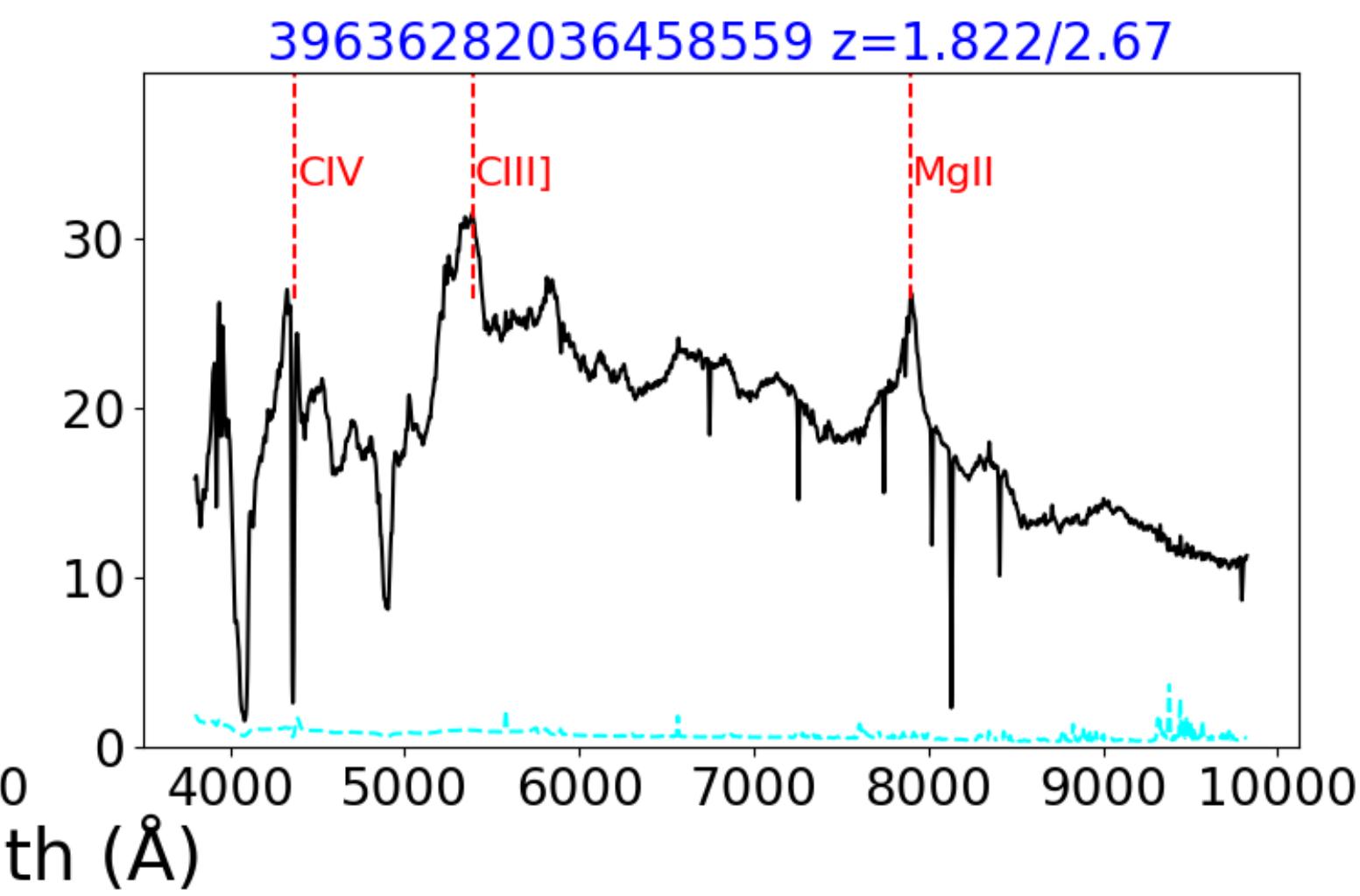
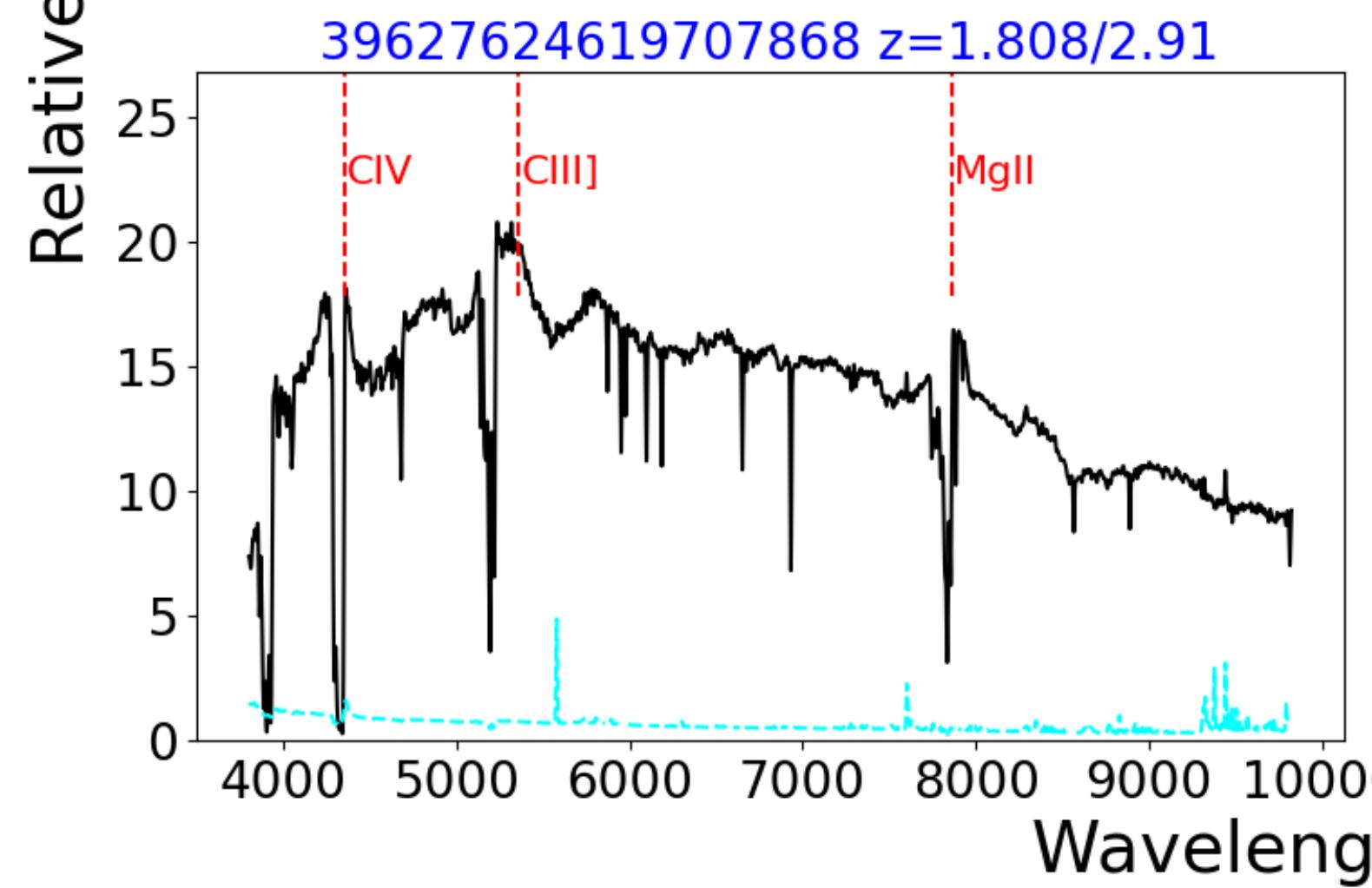
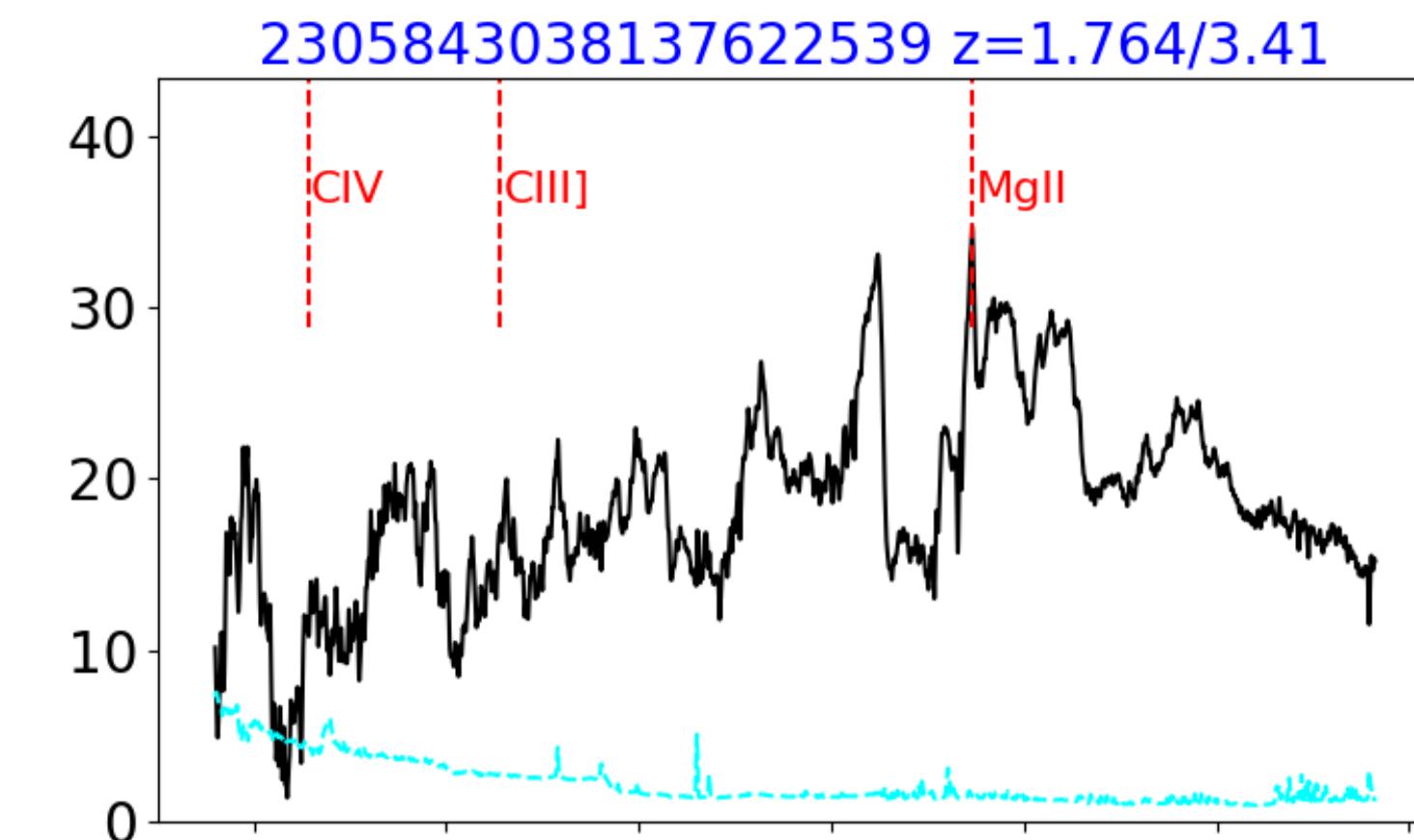
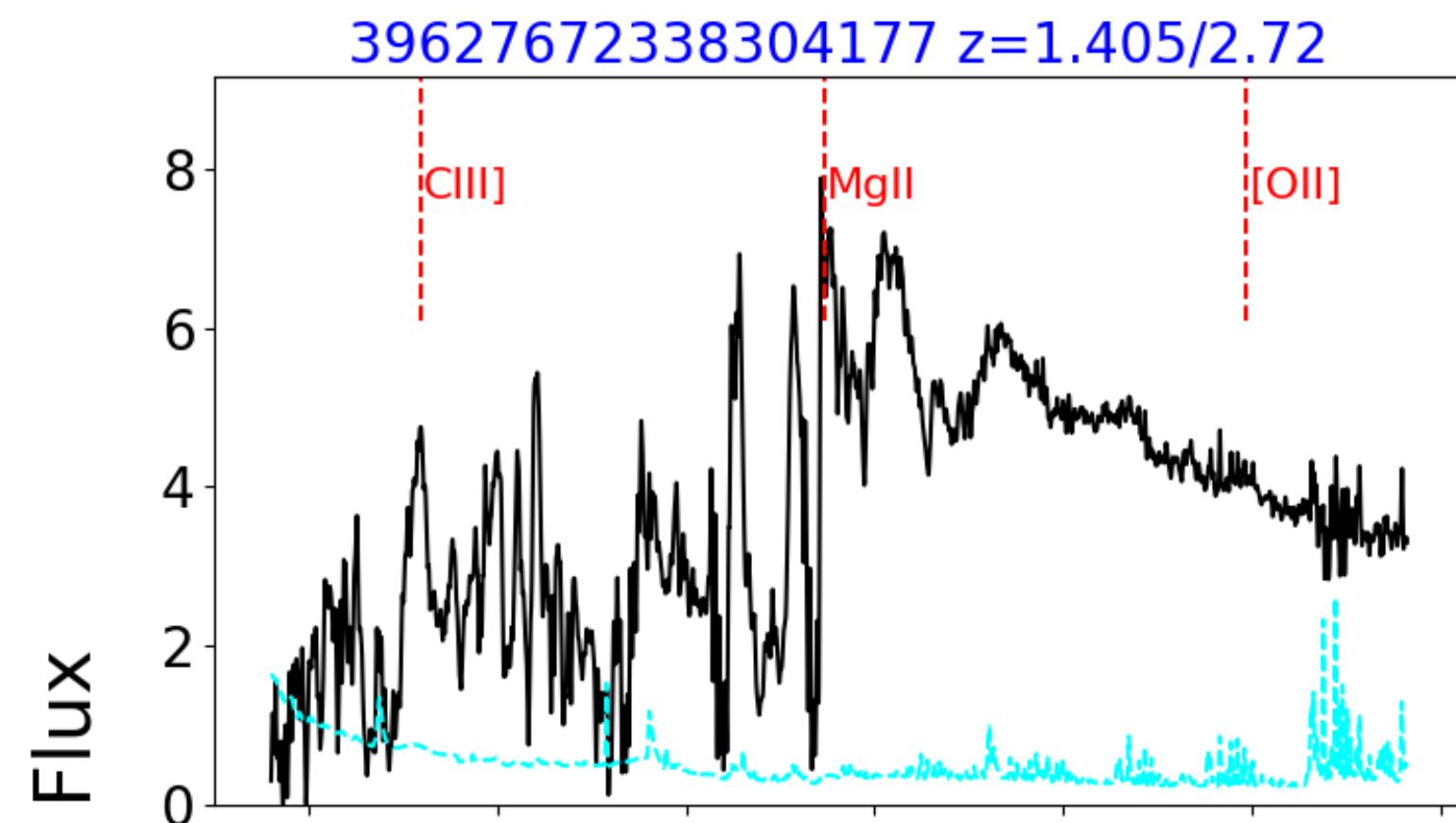
total int.: ~ 22 h

$\langle \text{SNR} \rangle$: ~ 120

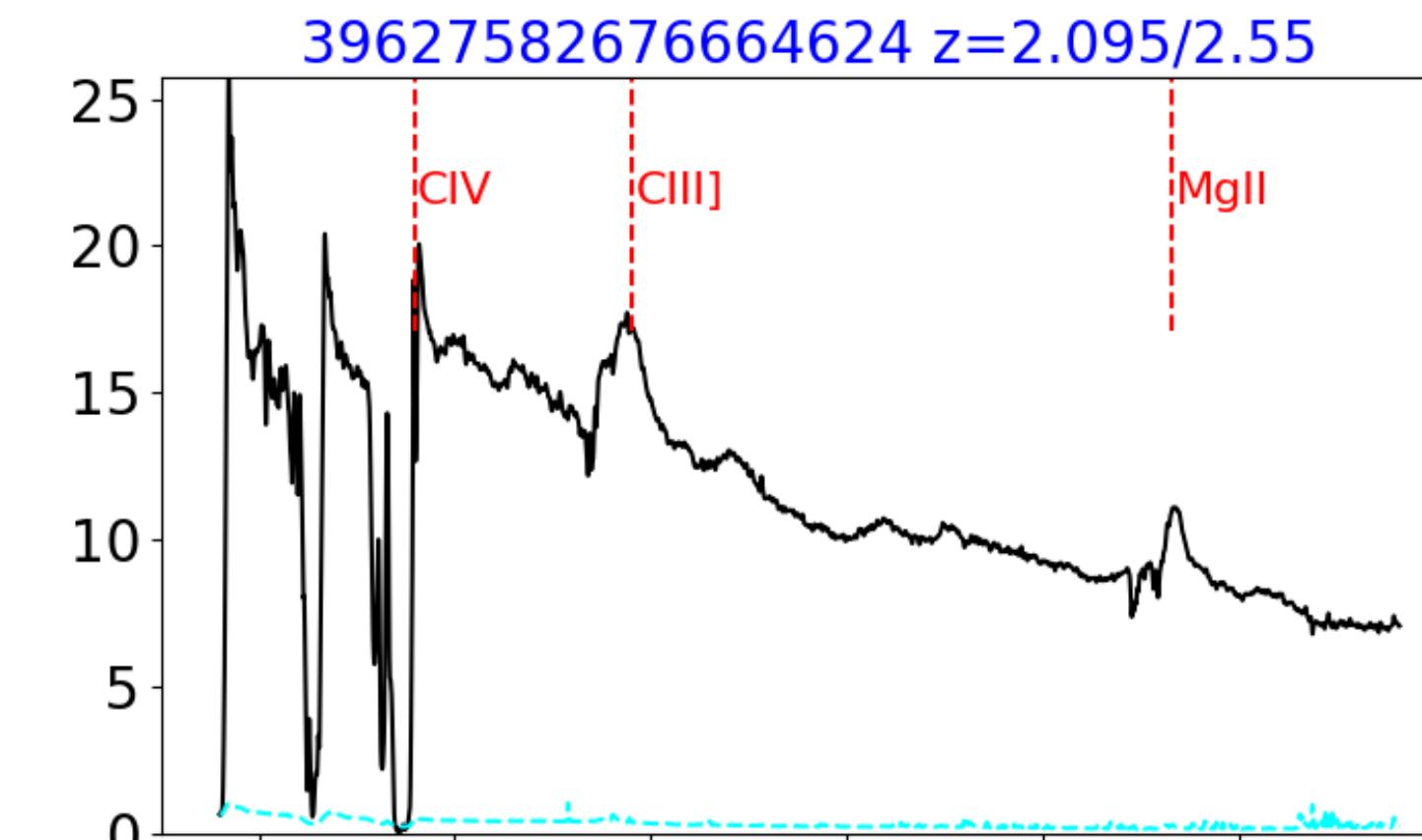
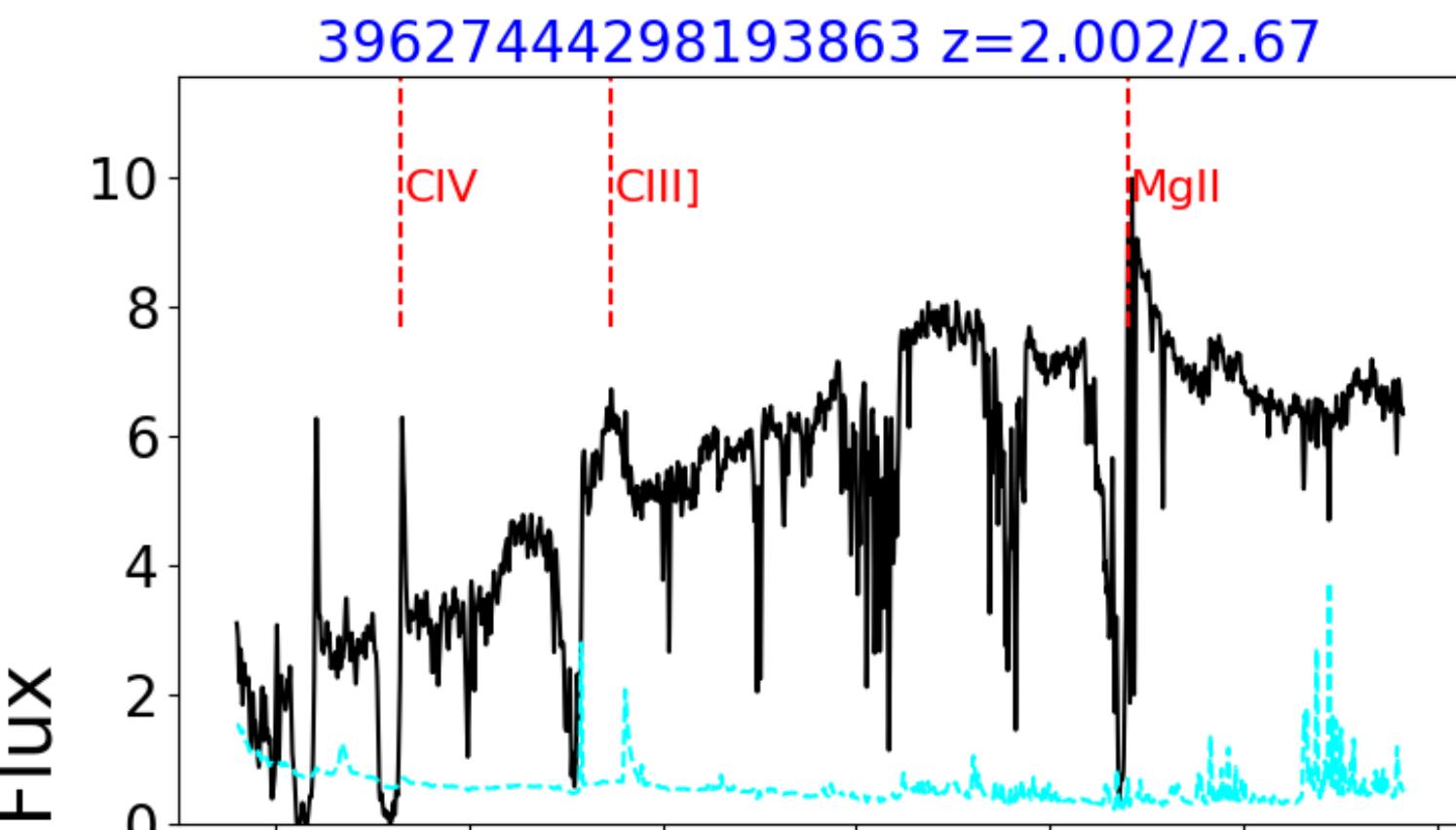
last obs.: 30/12/24



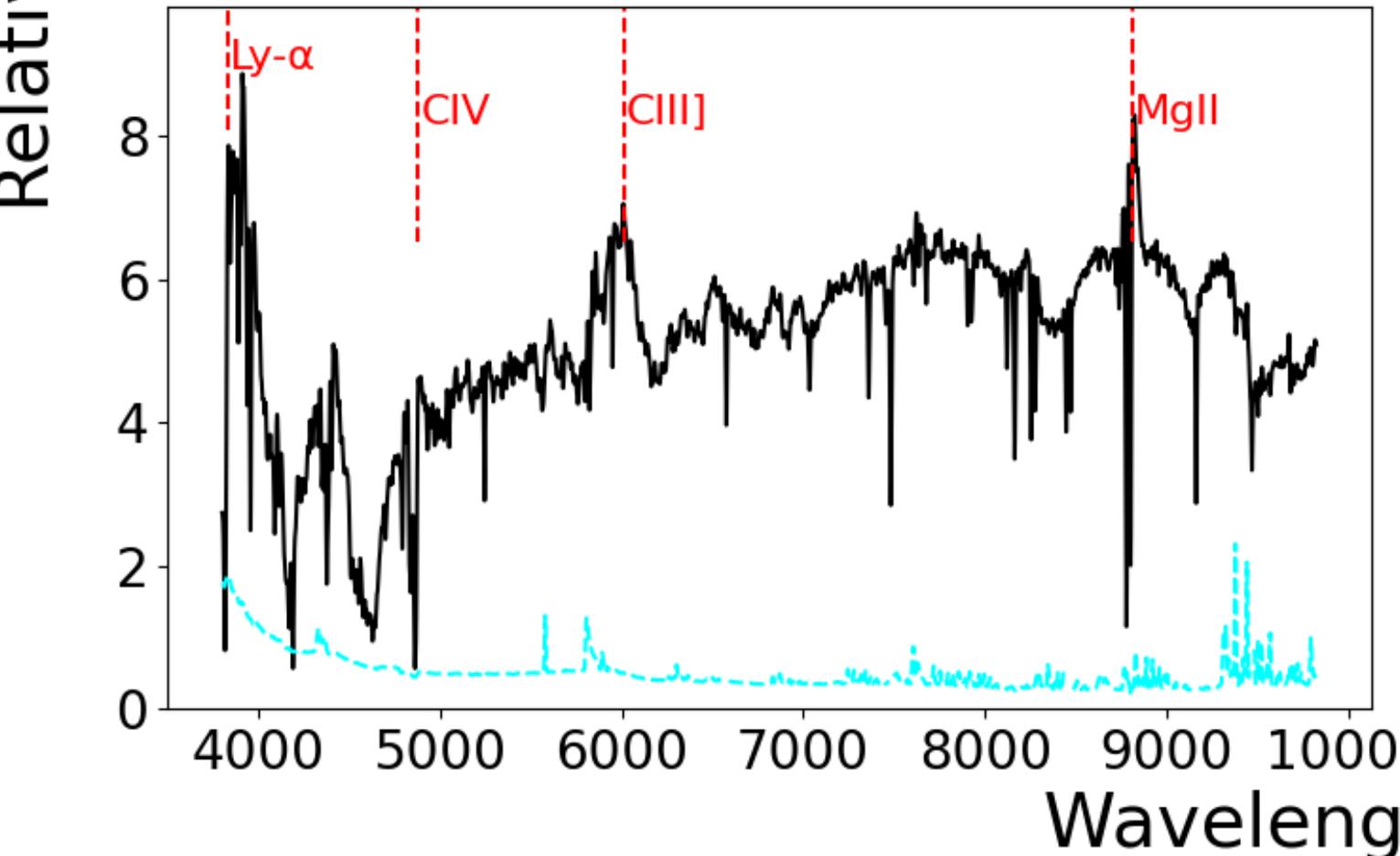
Rogues' Gallery



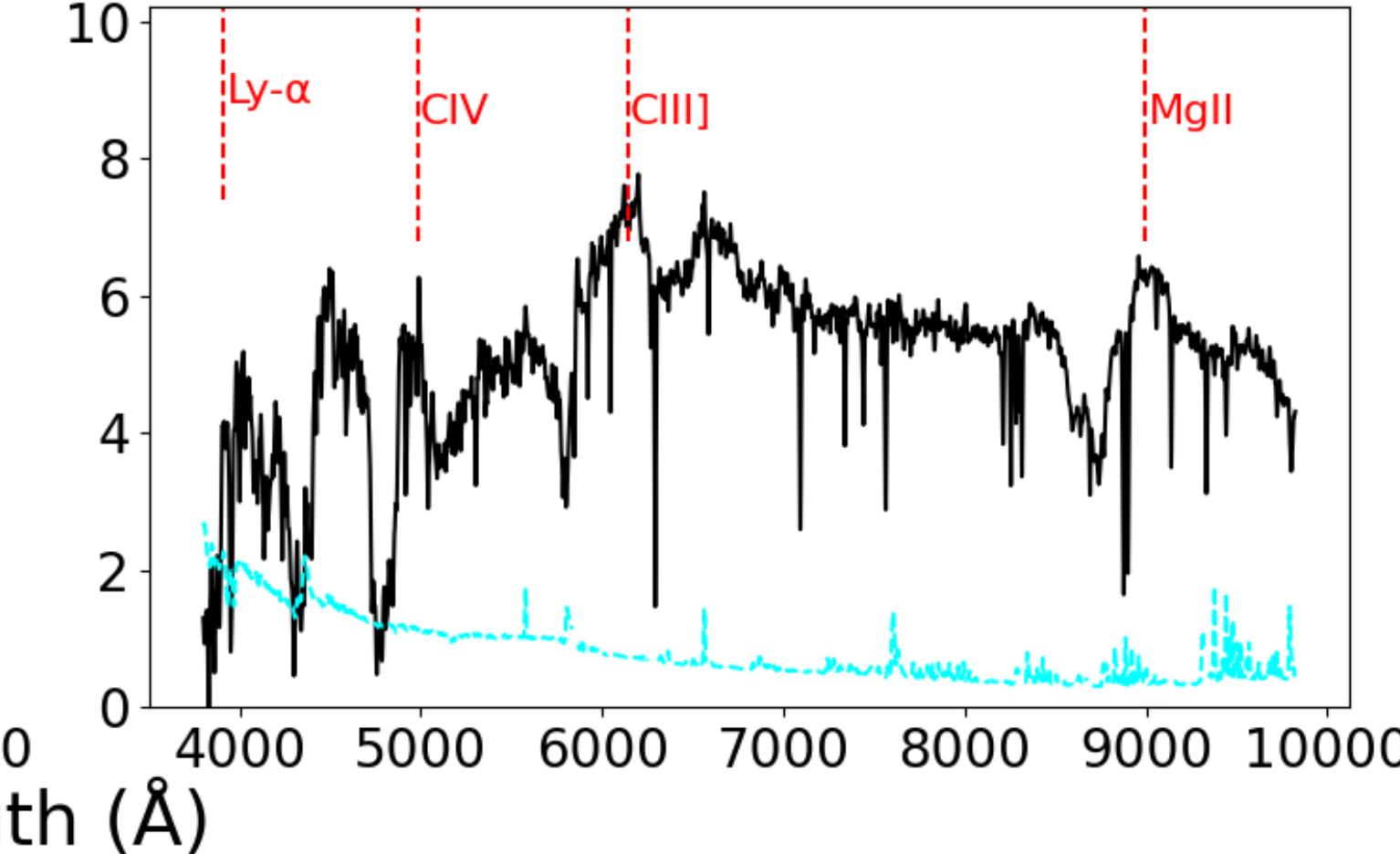
Relative Flux

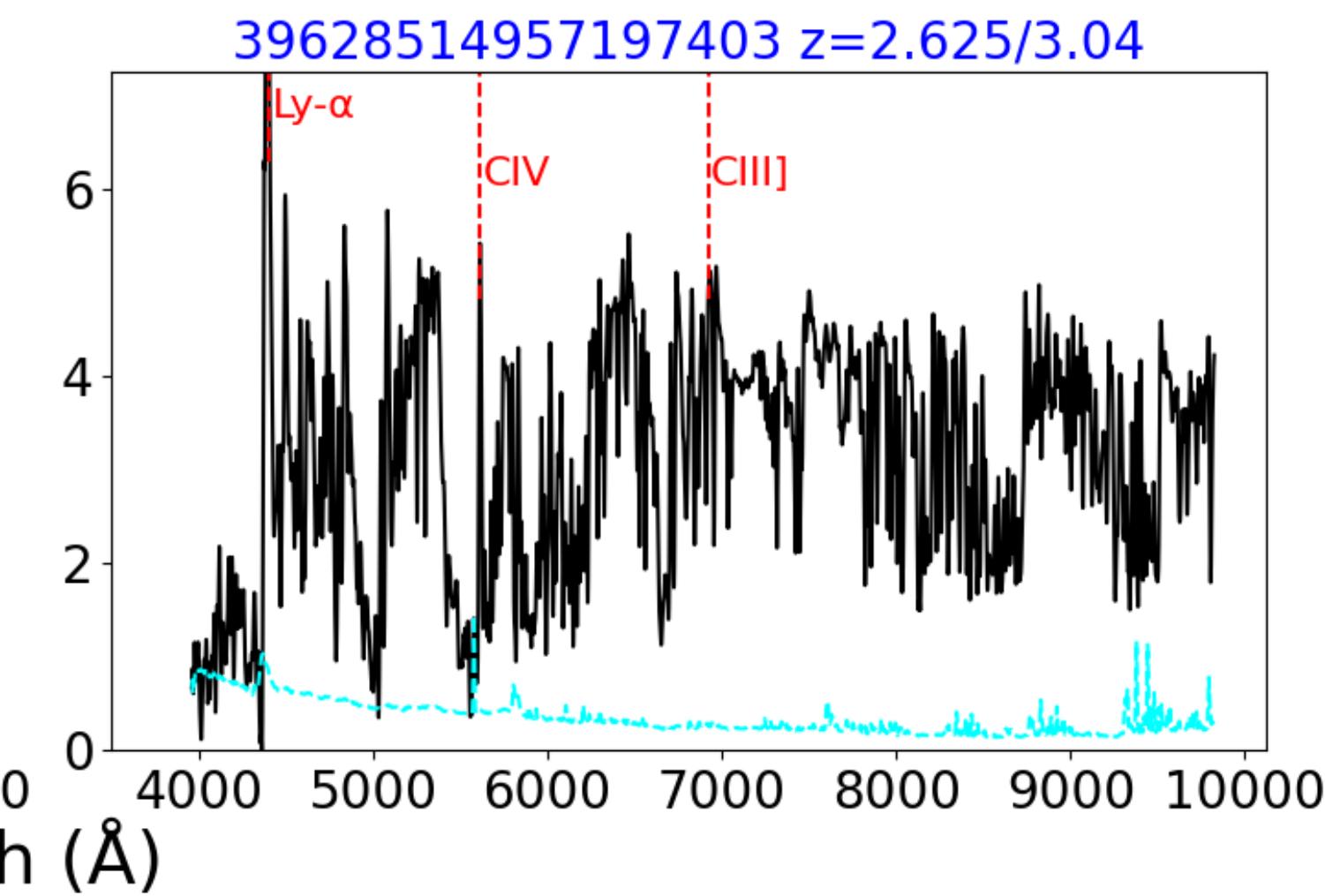
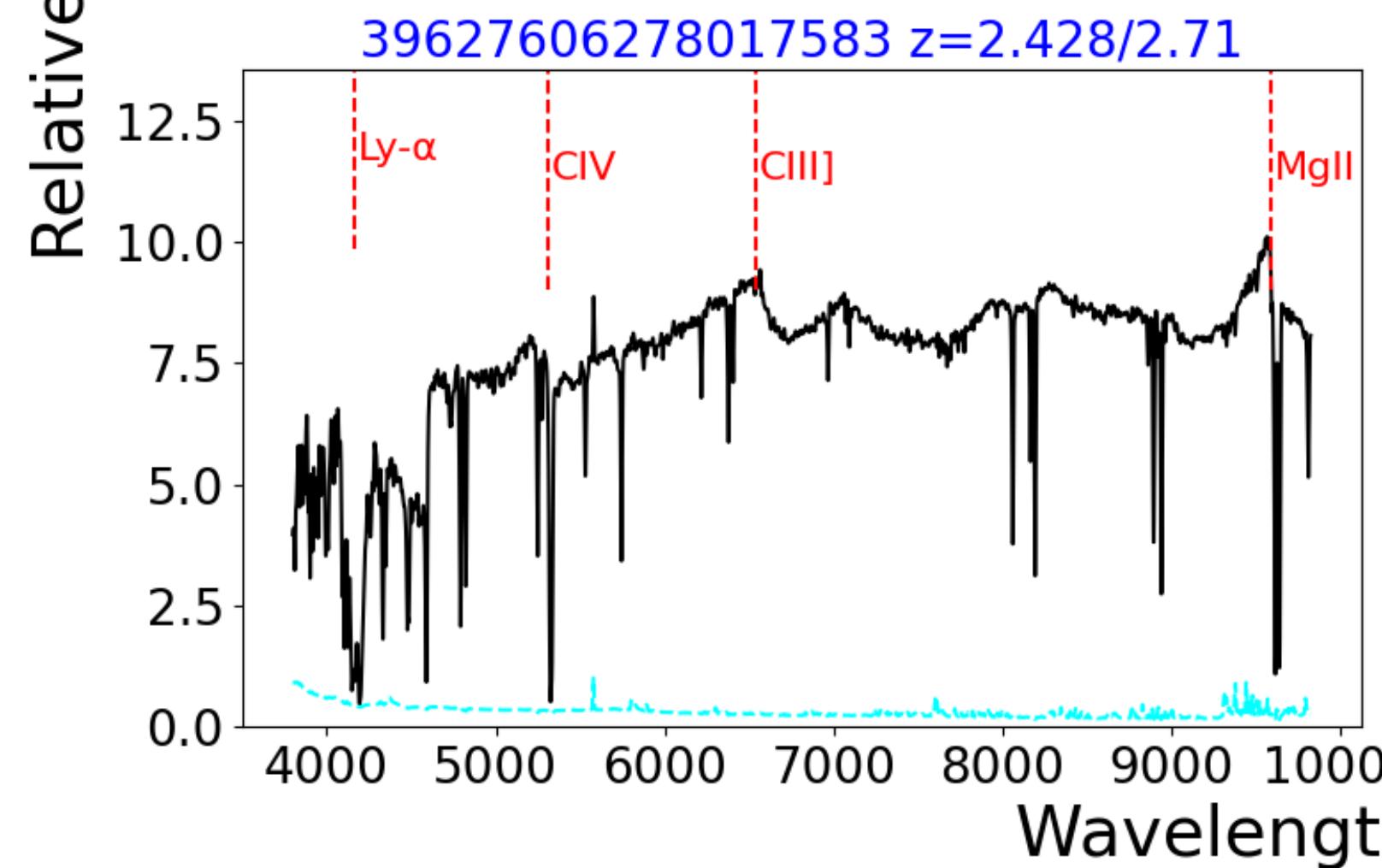
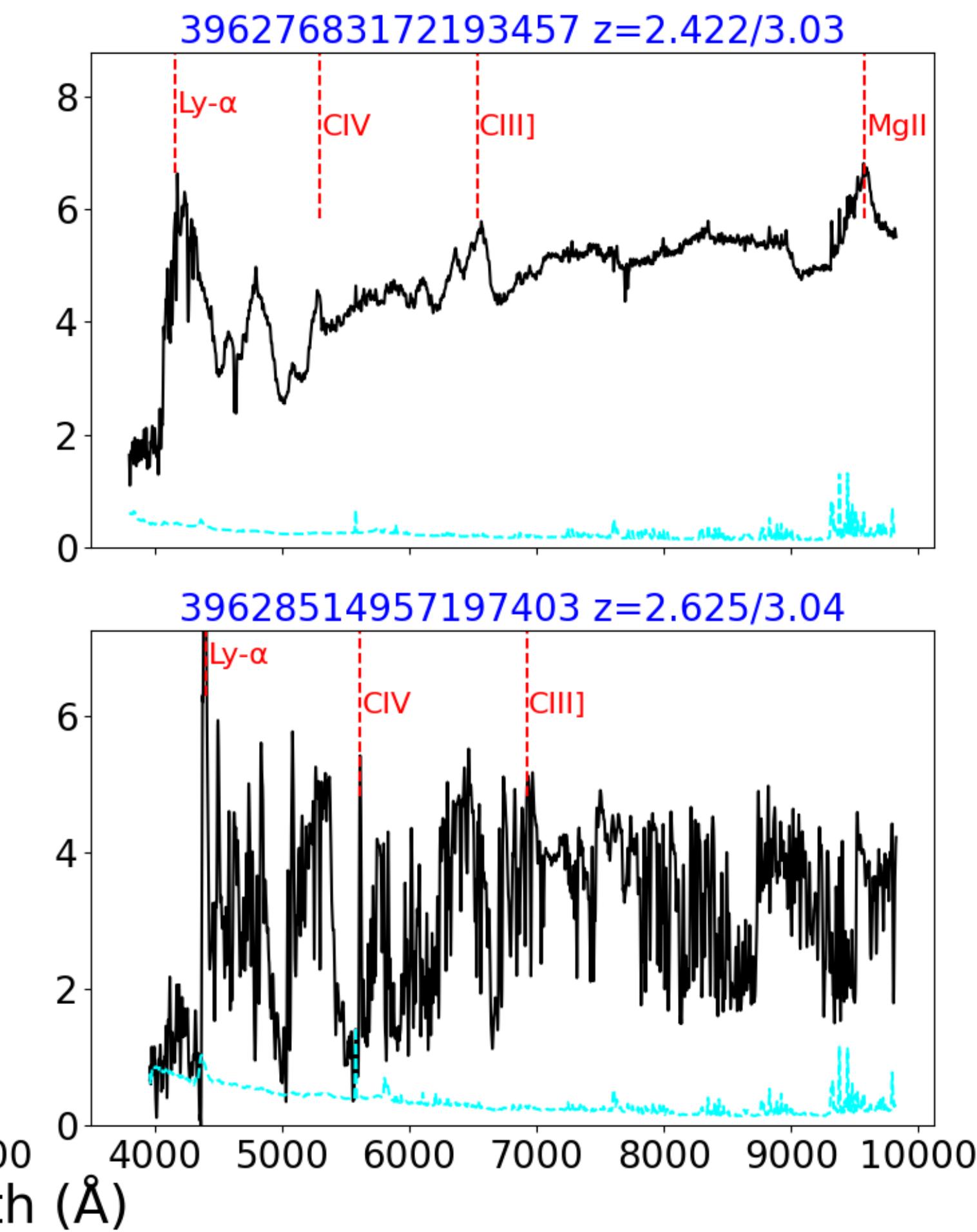
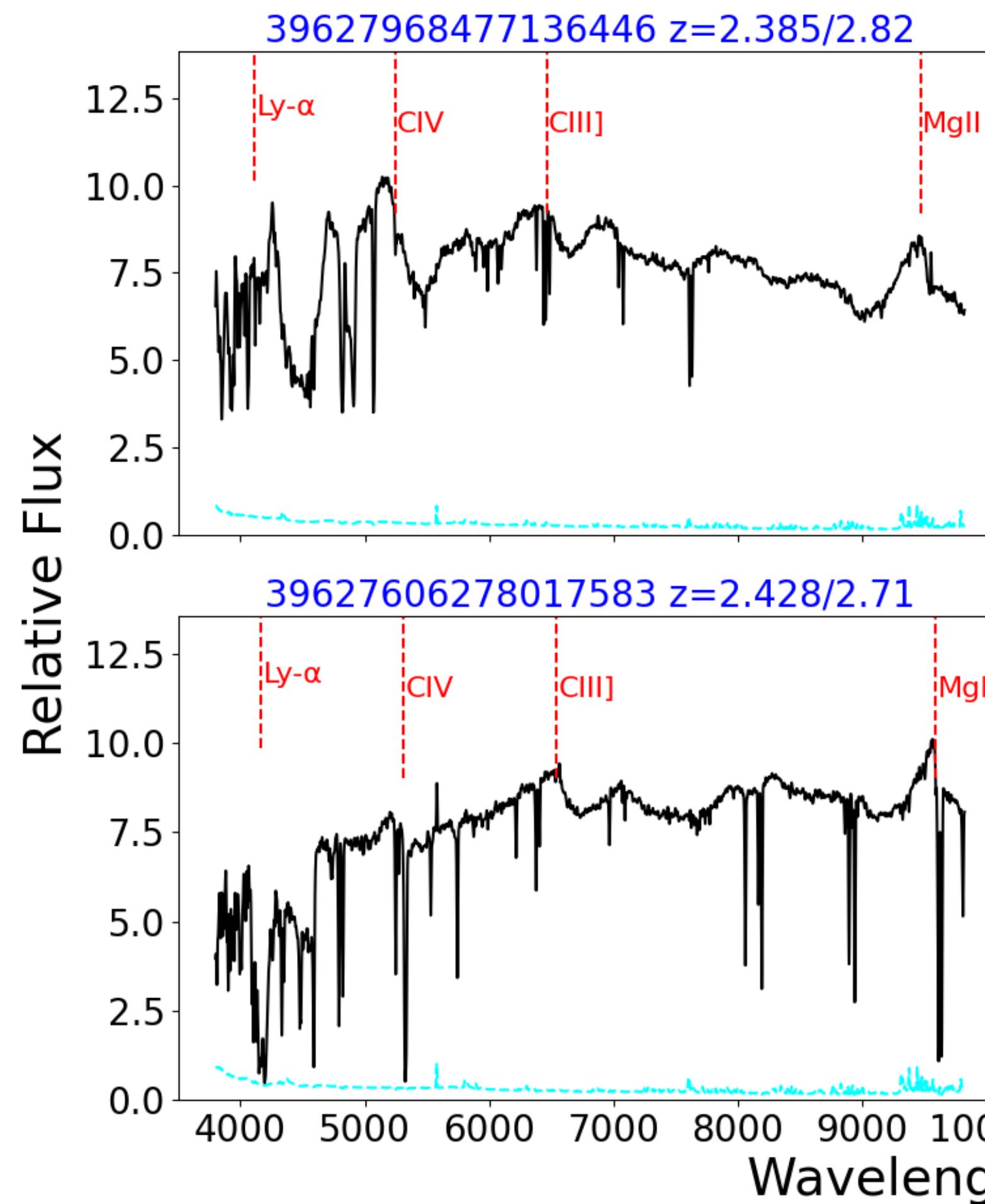


39627520894571719 z=2.148/2.65



39627450631593071 z=2.216/2.74

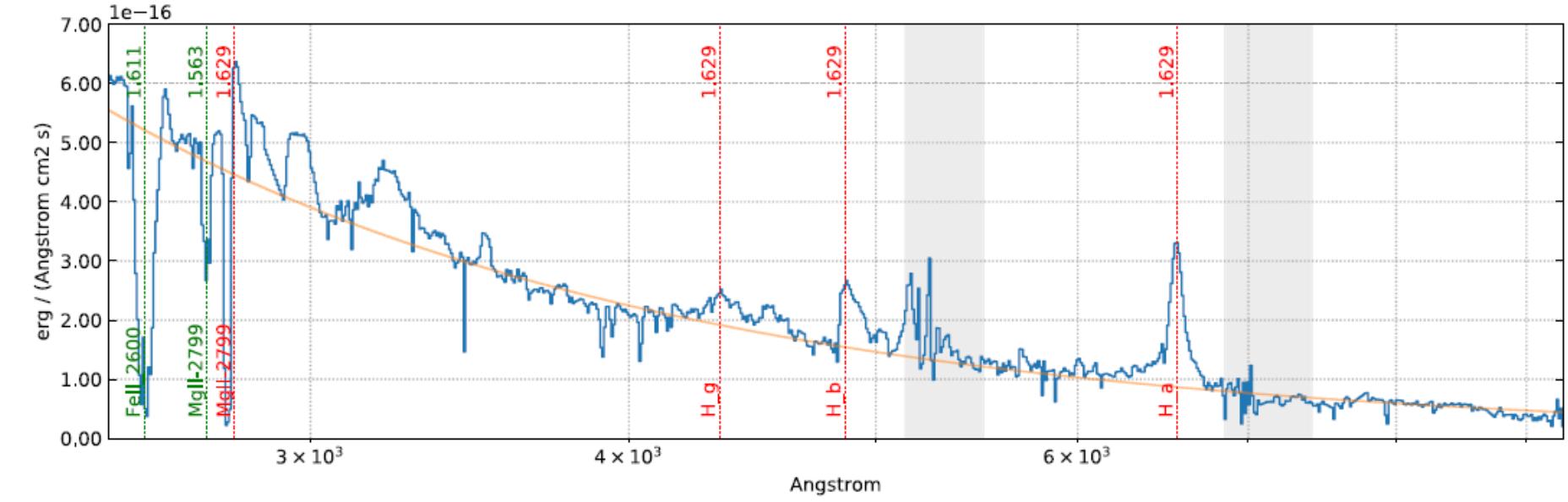
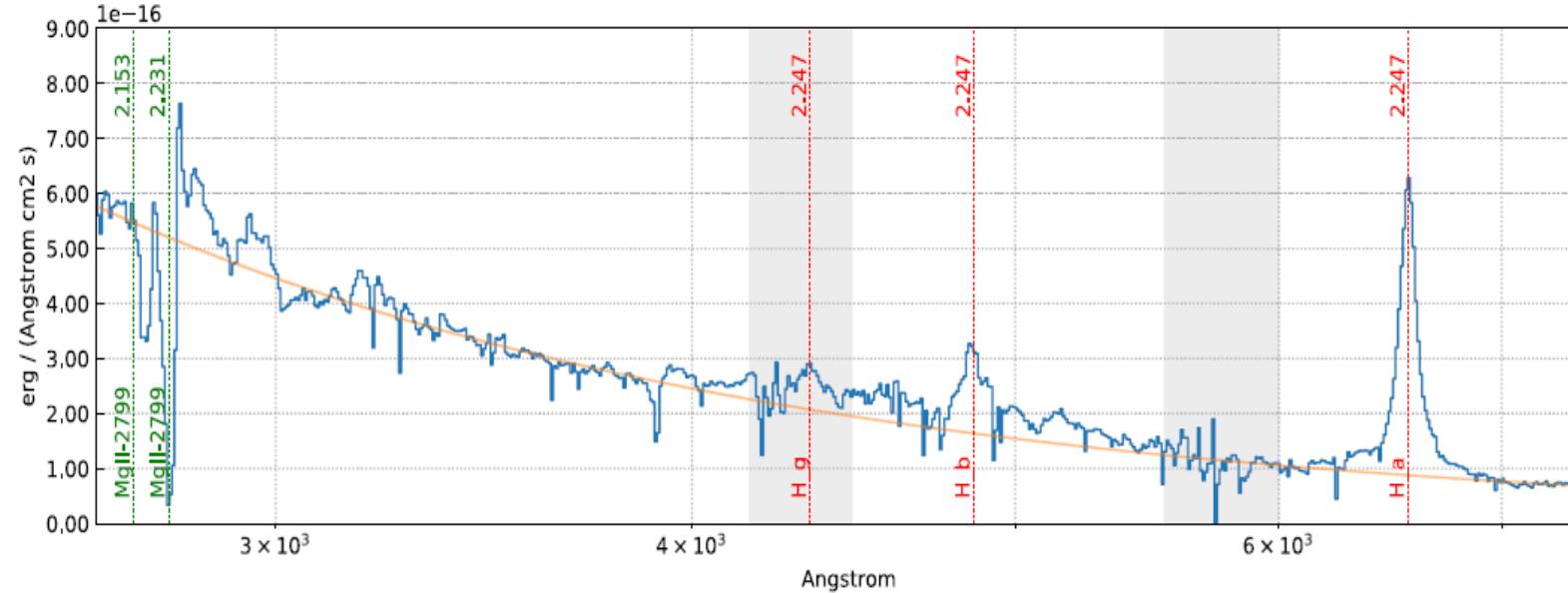
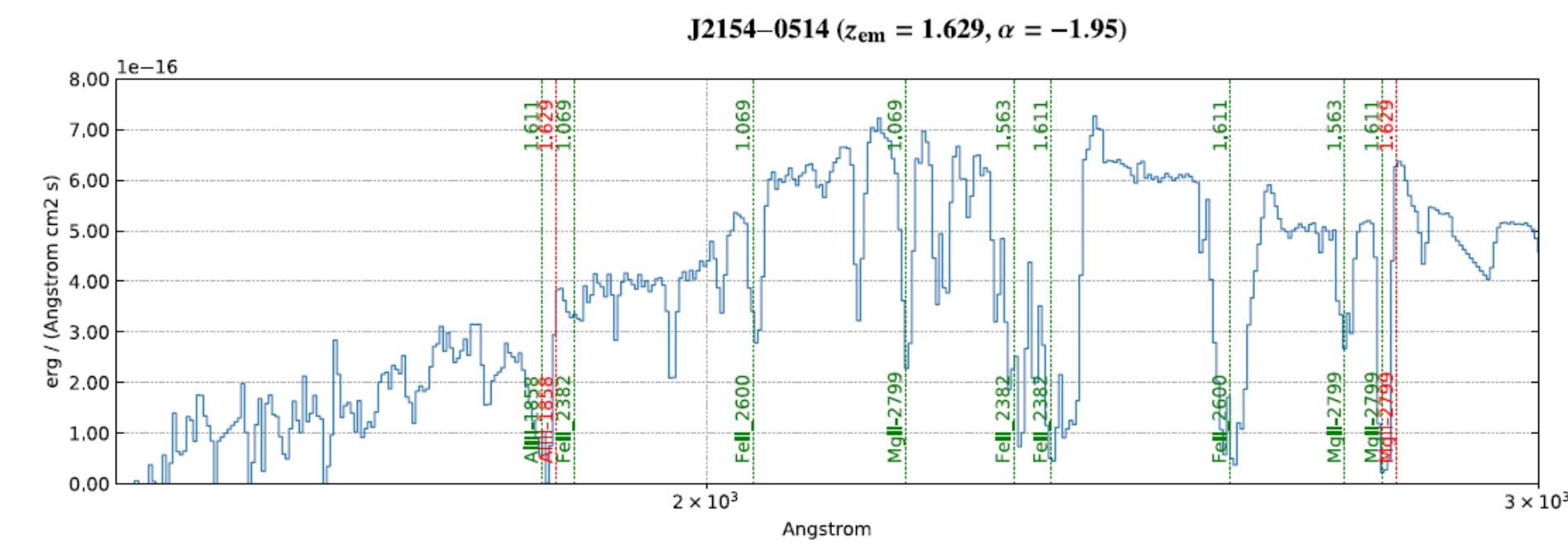
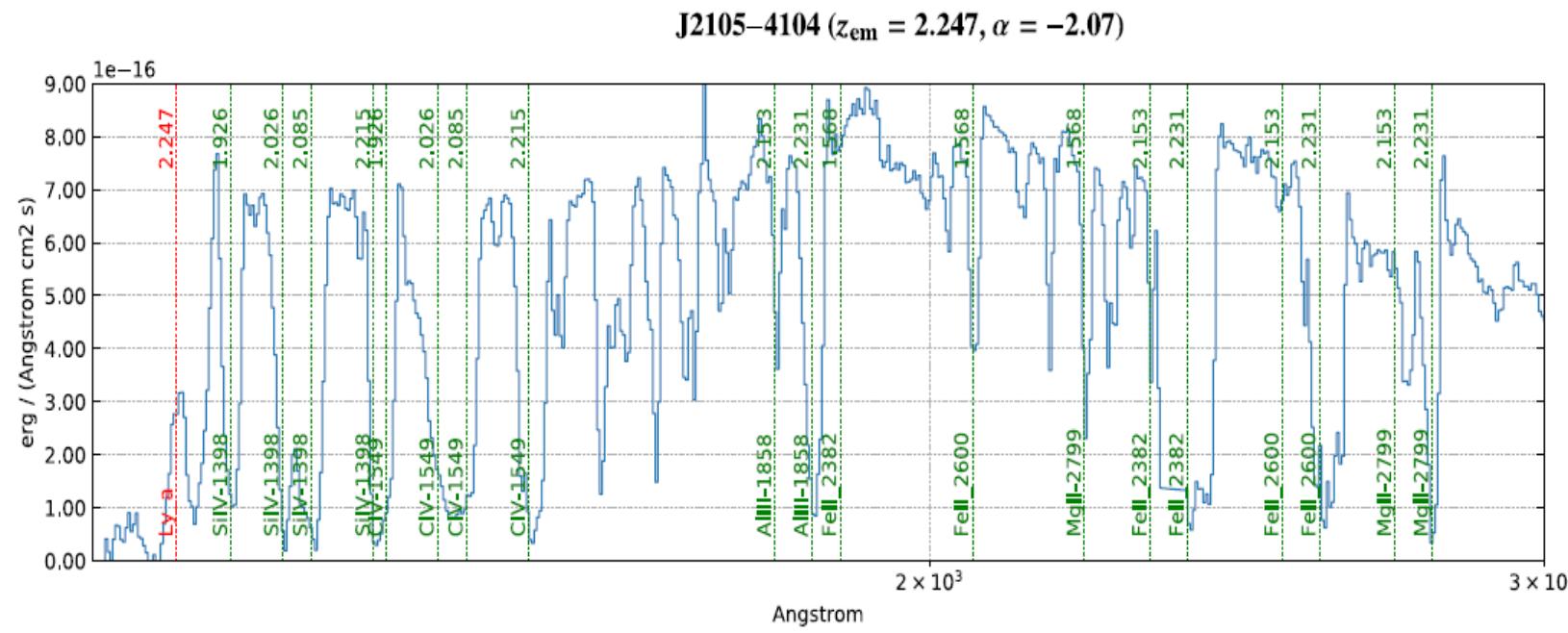




Near-infrared spectroscopy of extreme BAL QSOs from the QUBRICS bright quasar survey

Guido Cupani ^{b, 1,2★}, Giorgio Calderone ^{b, 1}, Pierluigi Selvelli, ¹ Stefano Cristiani, ^{1,2,3}
Konstantina Boutsia ^{b, 4}, Andrea Grazian, ⁵ Fabio Fontanot ^{b, 1,2} Francesco Guarneri ^{b, 1,6}
Valentina D’Odorico, ^{1,2,7} Emanuele Giallongo⁸ and Nicola Menci⁸

2022



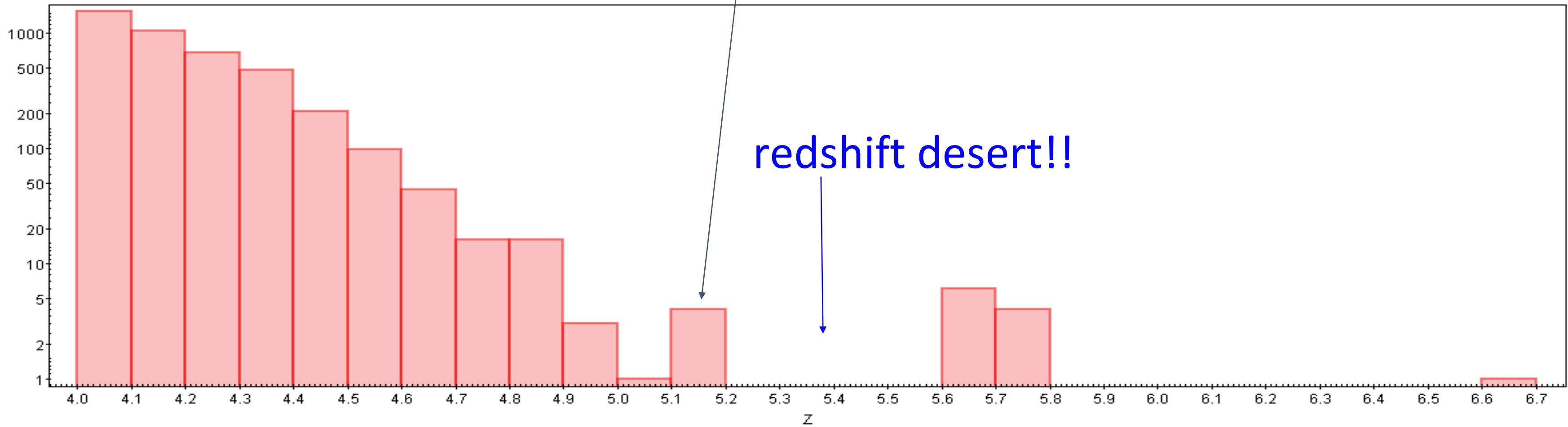
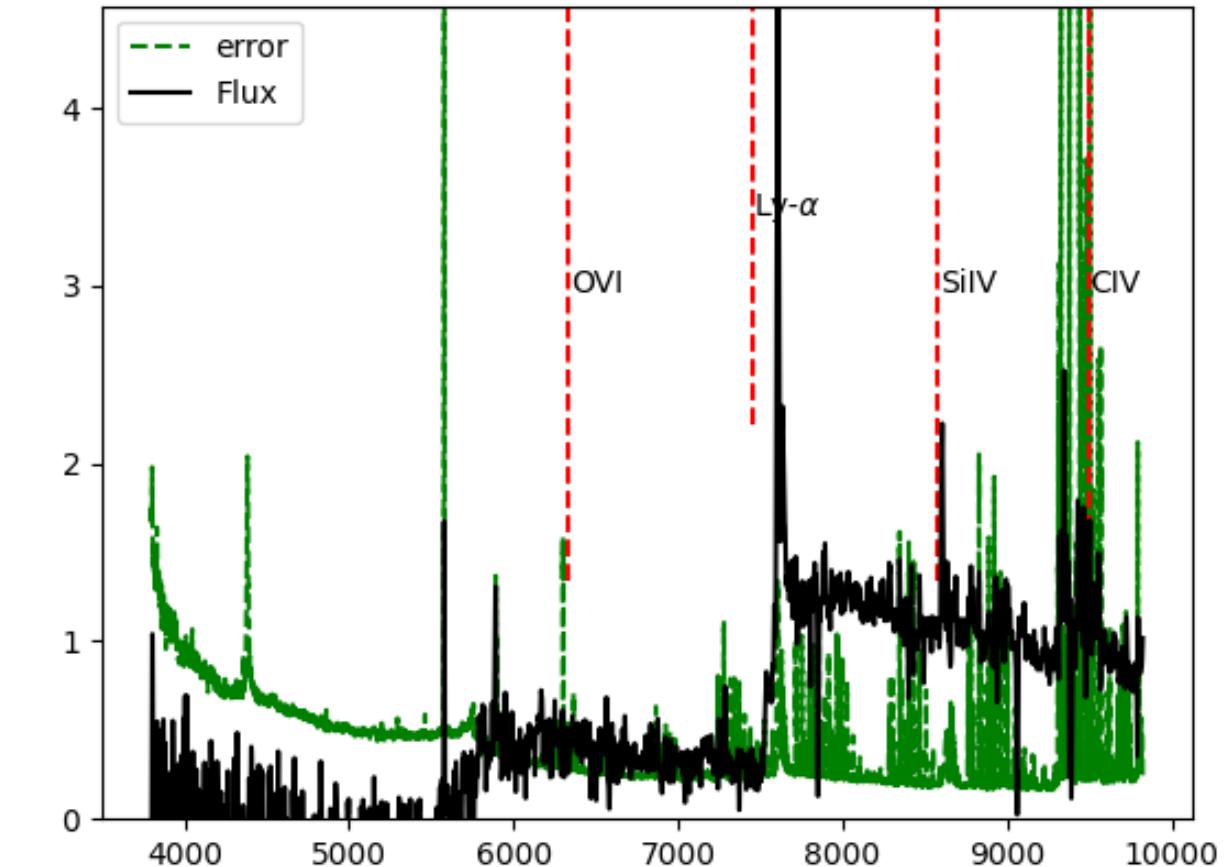
DESI DR1

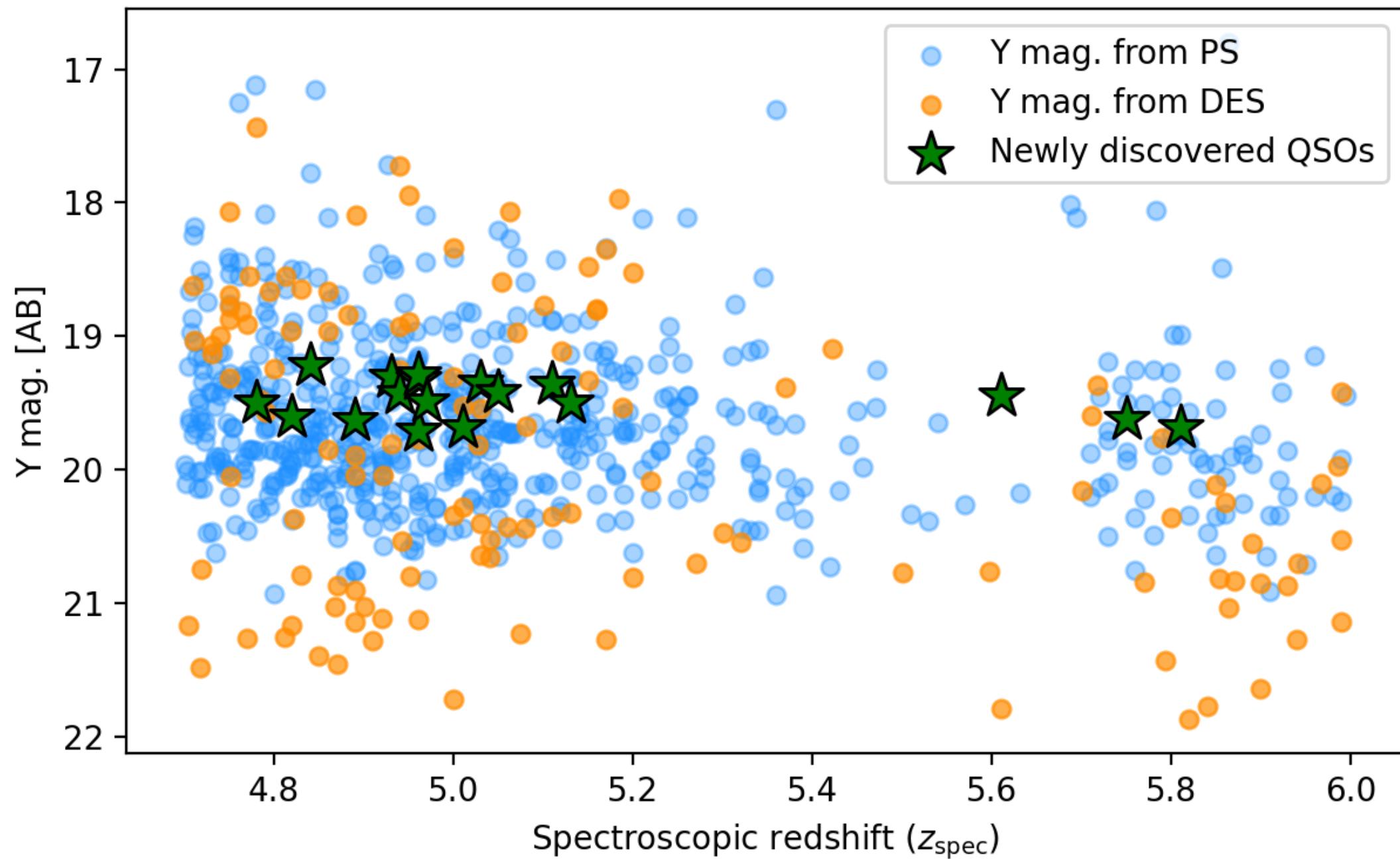
4171 QSOs

$z > 4$

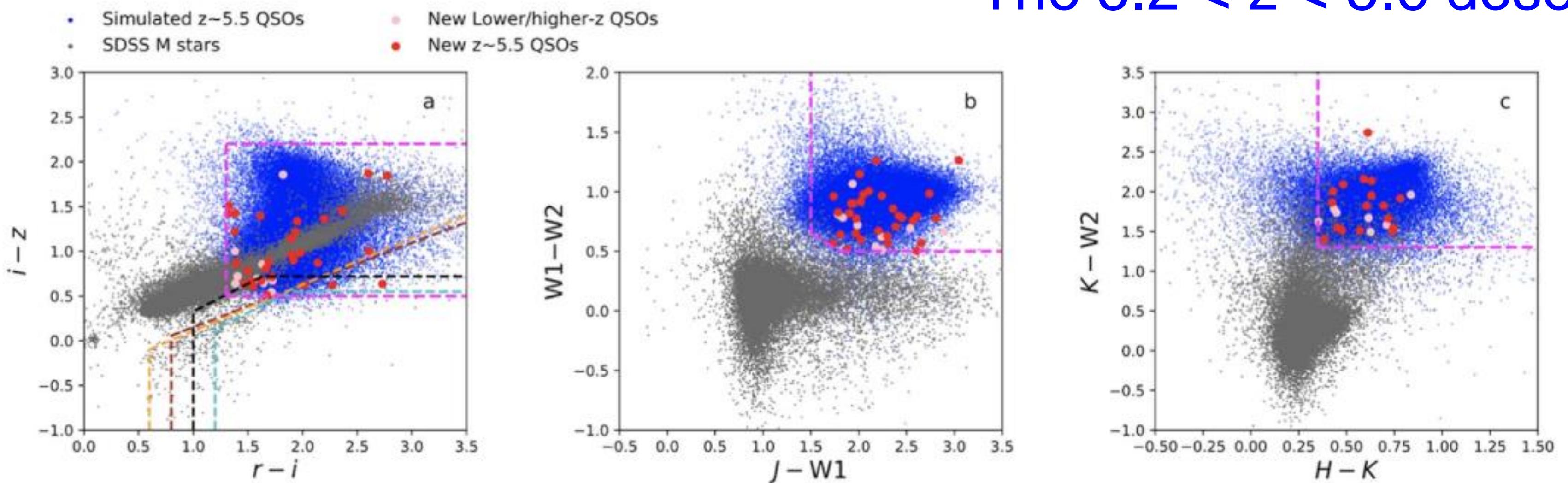
$\text{DEC} < 15$

highest z mag_z: 22.21 39627398114707701 z=5.126





The $5.2 < z < 5.6$ desert



THE ASTROPHYSICAL JOURNAL, 871:199 (17pp), 2019 February 1

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<https://doi.org/10.3847/1538-4357/aaf858>



Filling in the Quasar Redshift Gap at $z \sim 5.5$. II. A Complete Survey of Luminous Quasars in the Post-reionization Universe

Jinyi Yang^{1,2} , Feige Wang^{2,3} , Xiaohui Fan¹ , Xue-Bing Wu^{2,4} , Fuyan Bian⁵ , Eduardo Bañados⁶ , Minghao Yue¹ , Jan-Torge Schindler¹ , Qian Yang^{2,4} , Linhua Jiang² , Ian D. McGreer¹ , Richard Green¹, and Simon Dye⁷

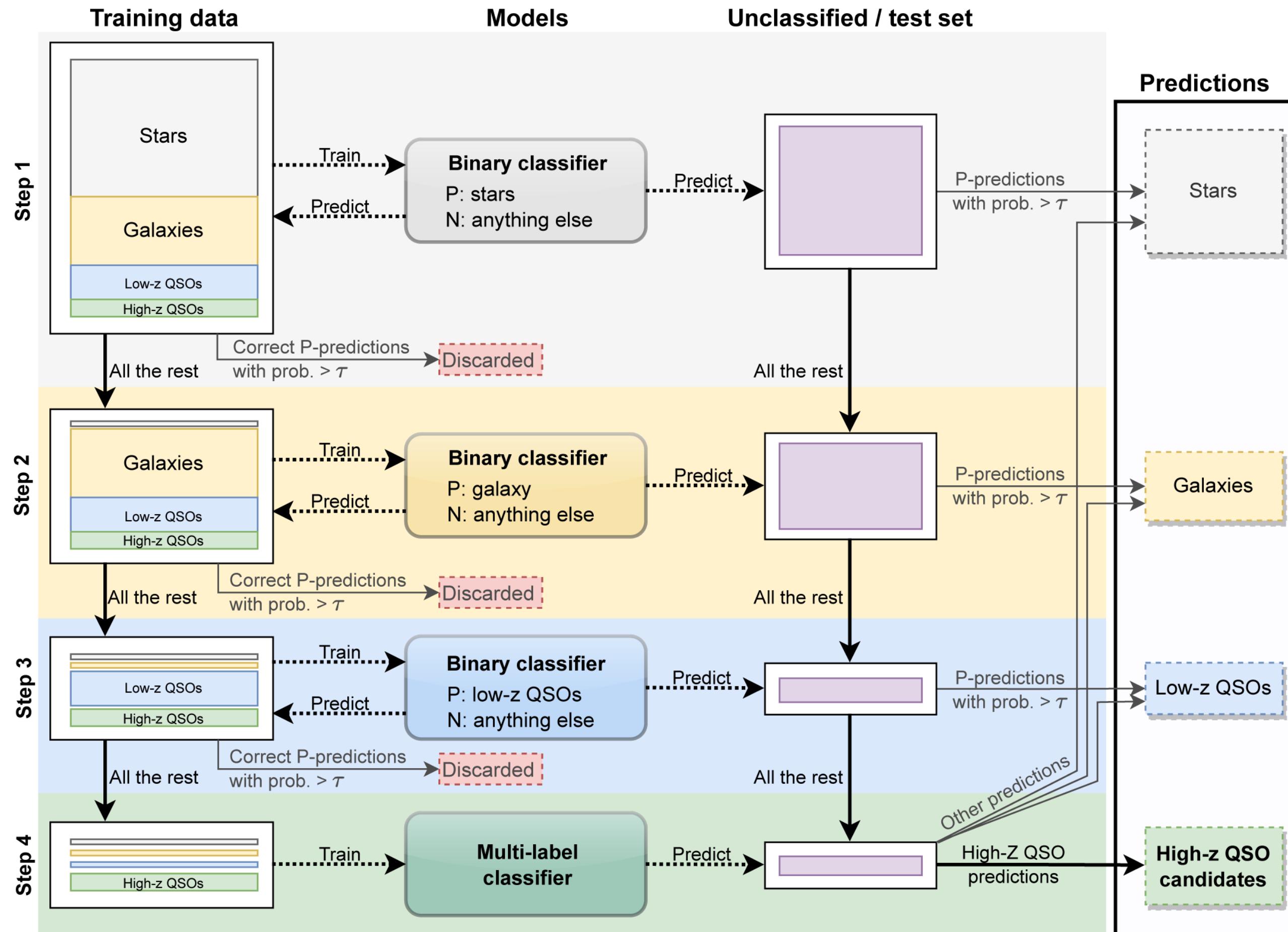
¹ Steward Observatory, University of Arizona, 933 N. Cherry Ave., Tucson, AZ, USA; jinyiyang@as.arizona.edu

² Kavli Institute for Astronomy and Astrophysics, Peking University, Beijing 100871, People's Republic of China

³ Department of Physics, University of California, Santa Barbara, CA 93106-9530, USA

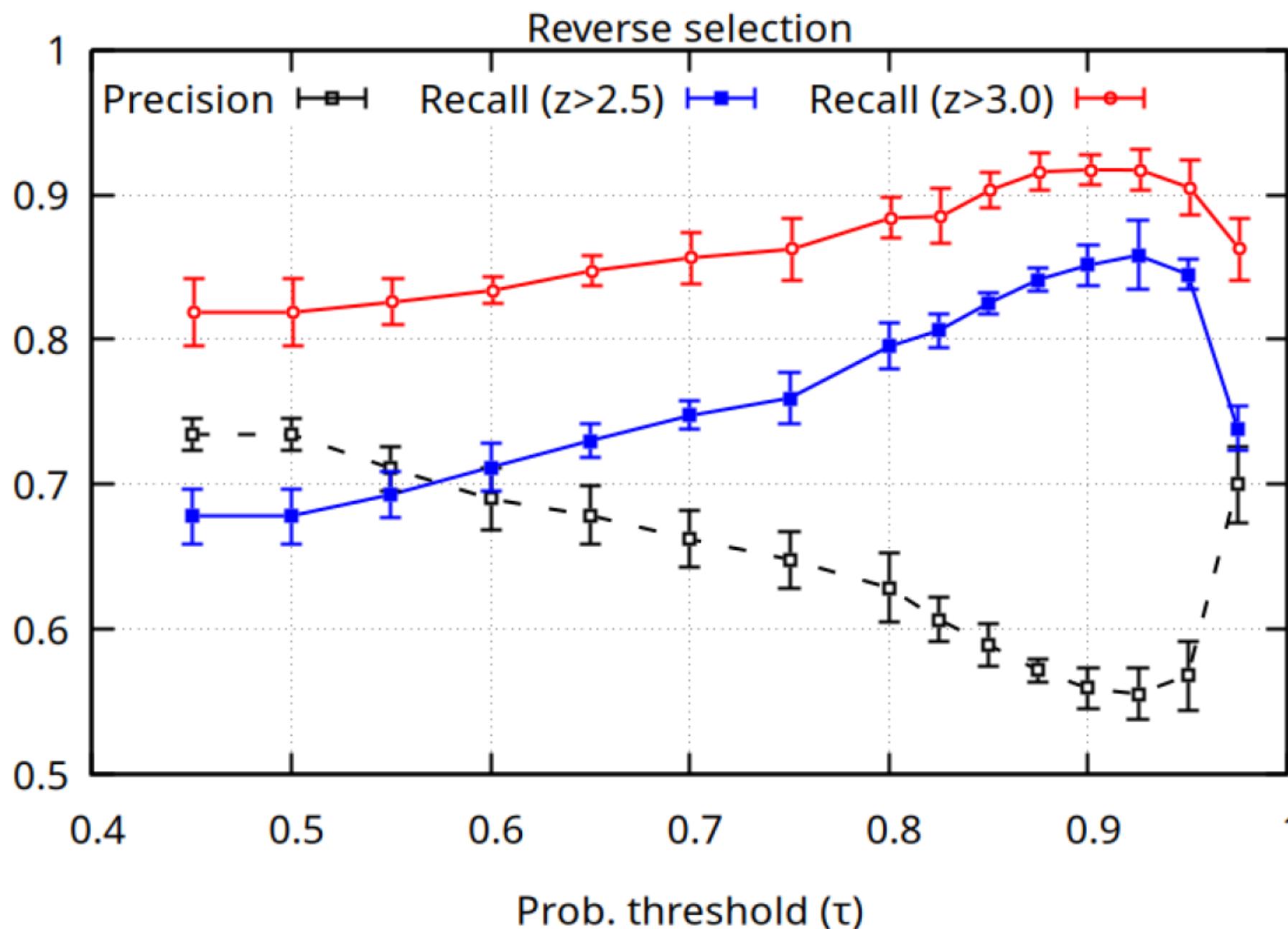
⁴ Department of Astronomy, School of Physics, Peking University, Beijing 100871, People's Republic of China

Reverse selection method XGB – Unbalanced data sets



Calderone+2024
A&A, 683, A34

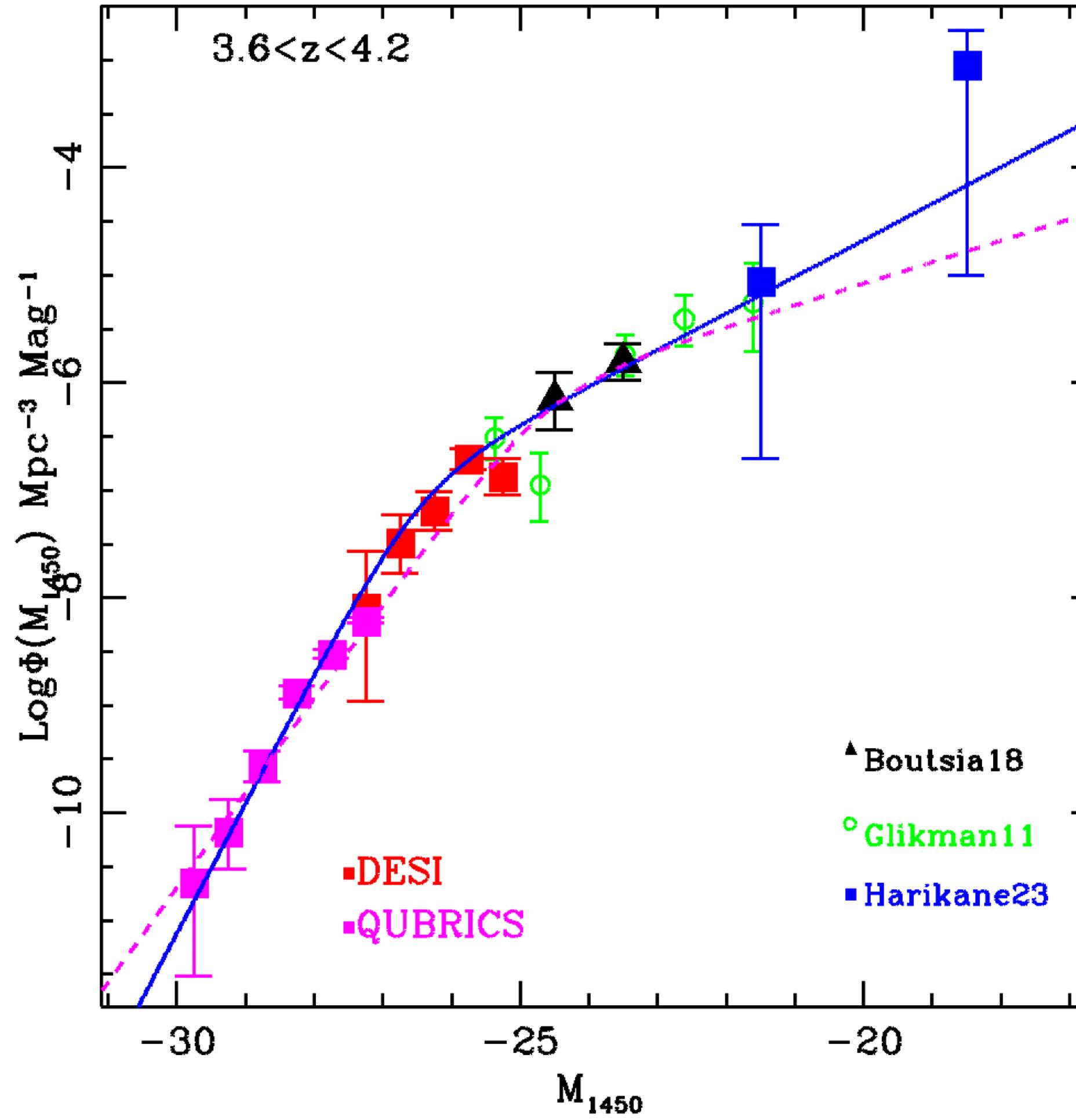
Calibrate probability threshold τ



Calderone+2024

- Increase recall, **slightly affect precision**;
- Optimal τ is the one maximizing recall;
- The better performances depend on the test set choice?
- ...run 100 analysis with random train/test splits, estimate recall at $z > 2.5, 3$.

SCIENTIFIC RESULTS



Boutsia+2025 in prep.

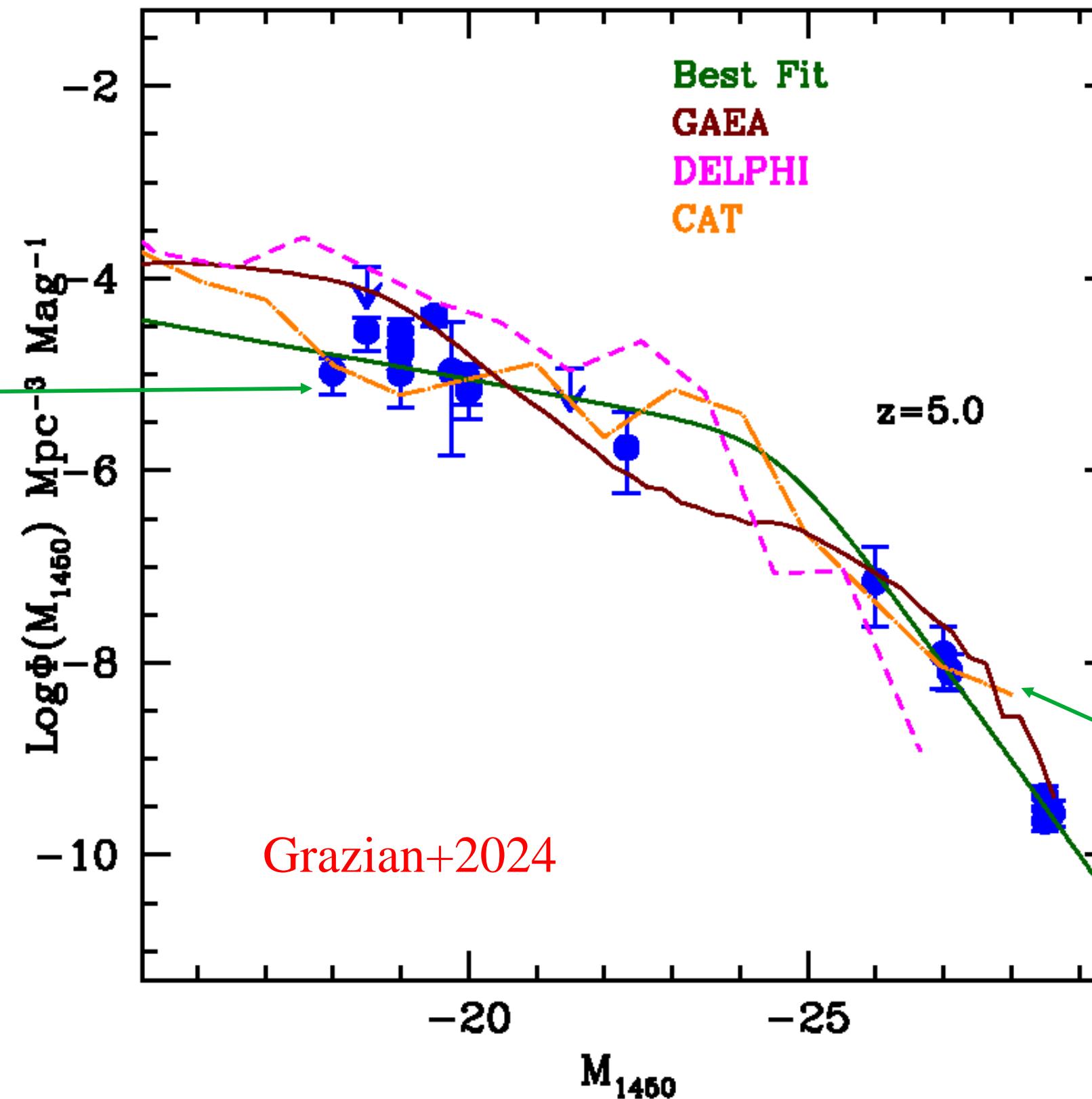
Table 3. The Best-Fit Parameters of the QSO LF at $3.6 \leq z \leq 4.2$ in the QUBRICS Footprint

α	β	M_{1450}^*	$\text{Log } \Phi^*$
$-1.850^{+0.150}_{-0.250}$	$-4.025^{+0.575}_{-0.425}$	$-26.50^{+0.85}_{-0.60}$	$-6.85^{+0.60}_{-0.45}$

Boutsia+2021

The $z \sim 5$ AGN Luminosity Function

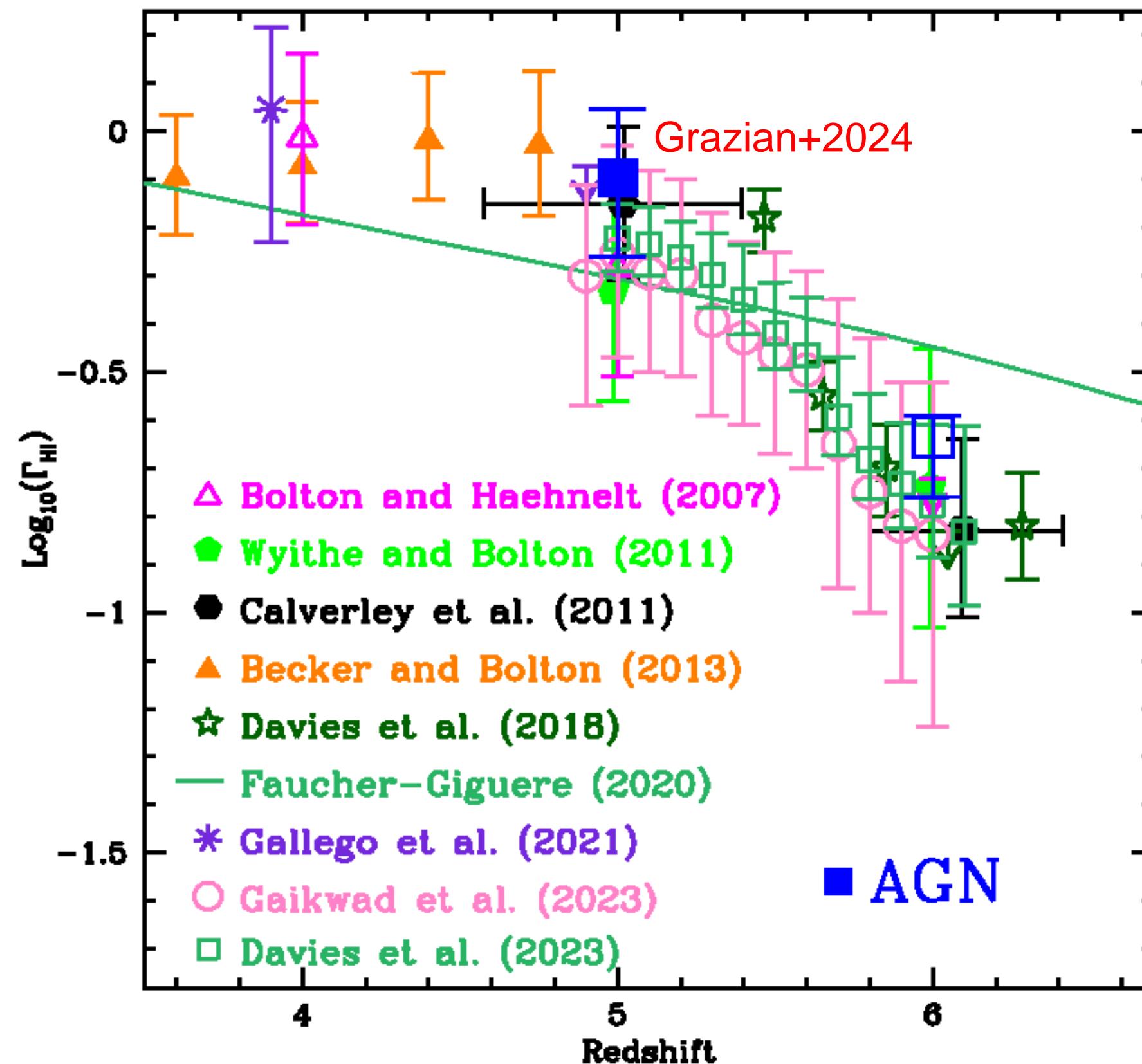
JWST:
Kocevski+2023
Harikane+2023
Matthee+2024
Greene+2024



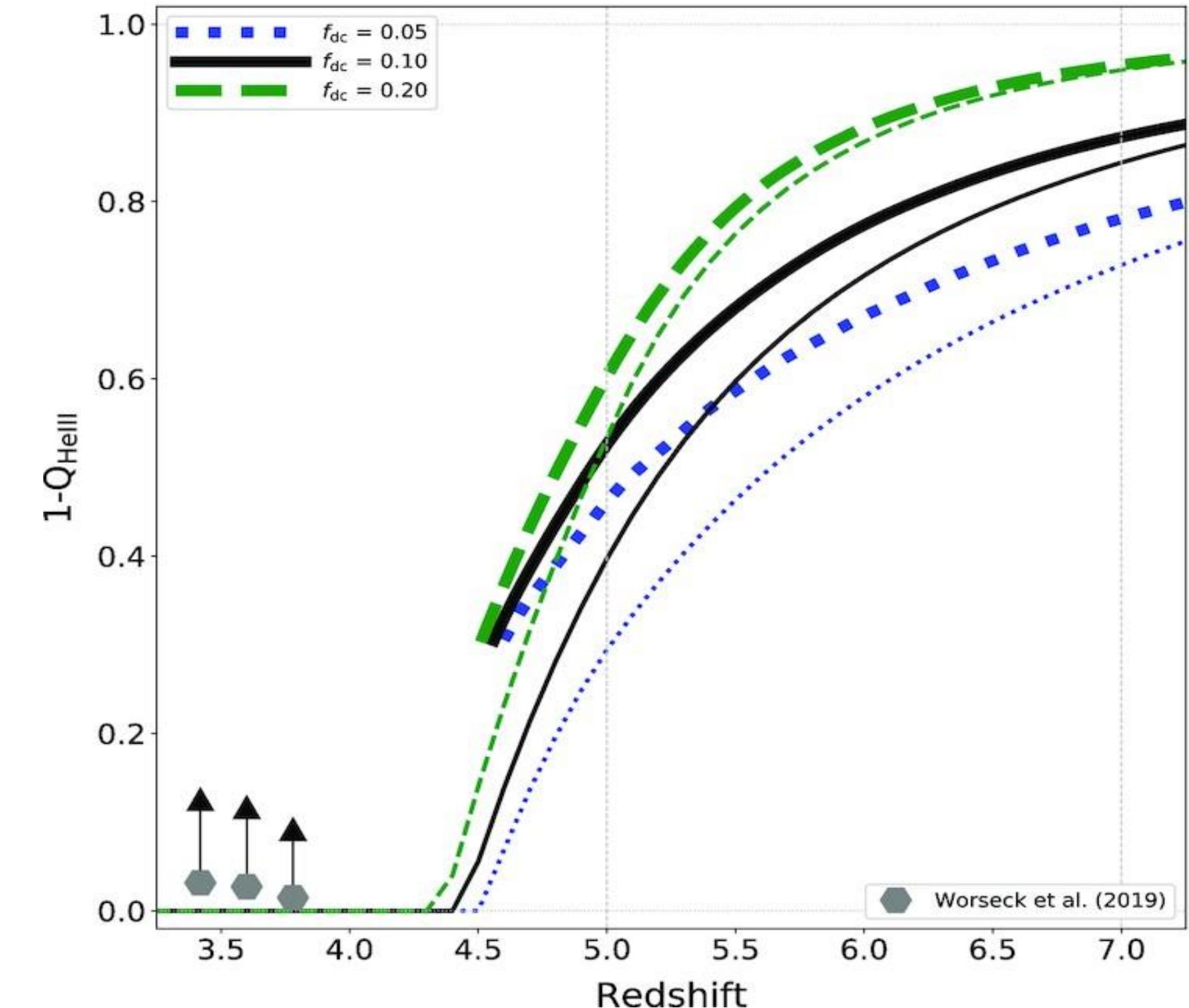
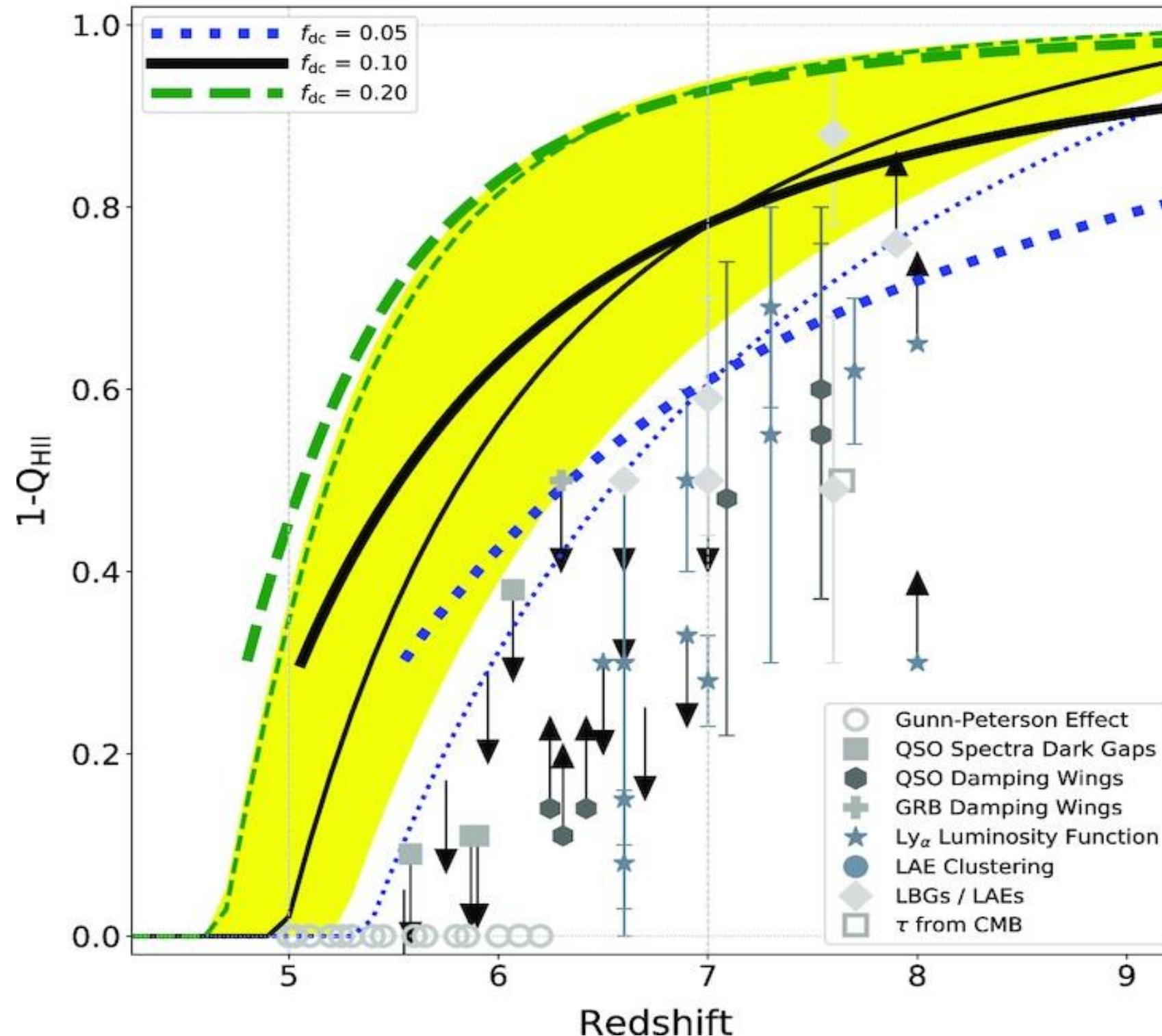
RUBICON (Reionizing
the Universe with
Bright COsmological
Nuclei)
Grazian+2023

QUBRICS
and
RUBICON
Surveys

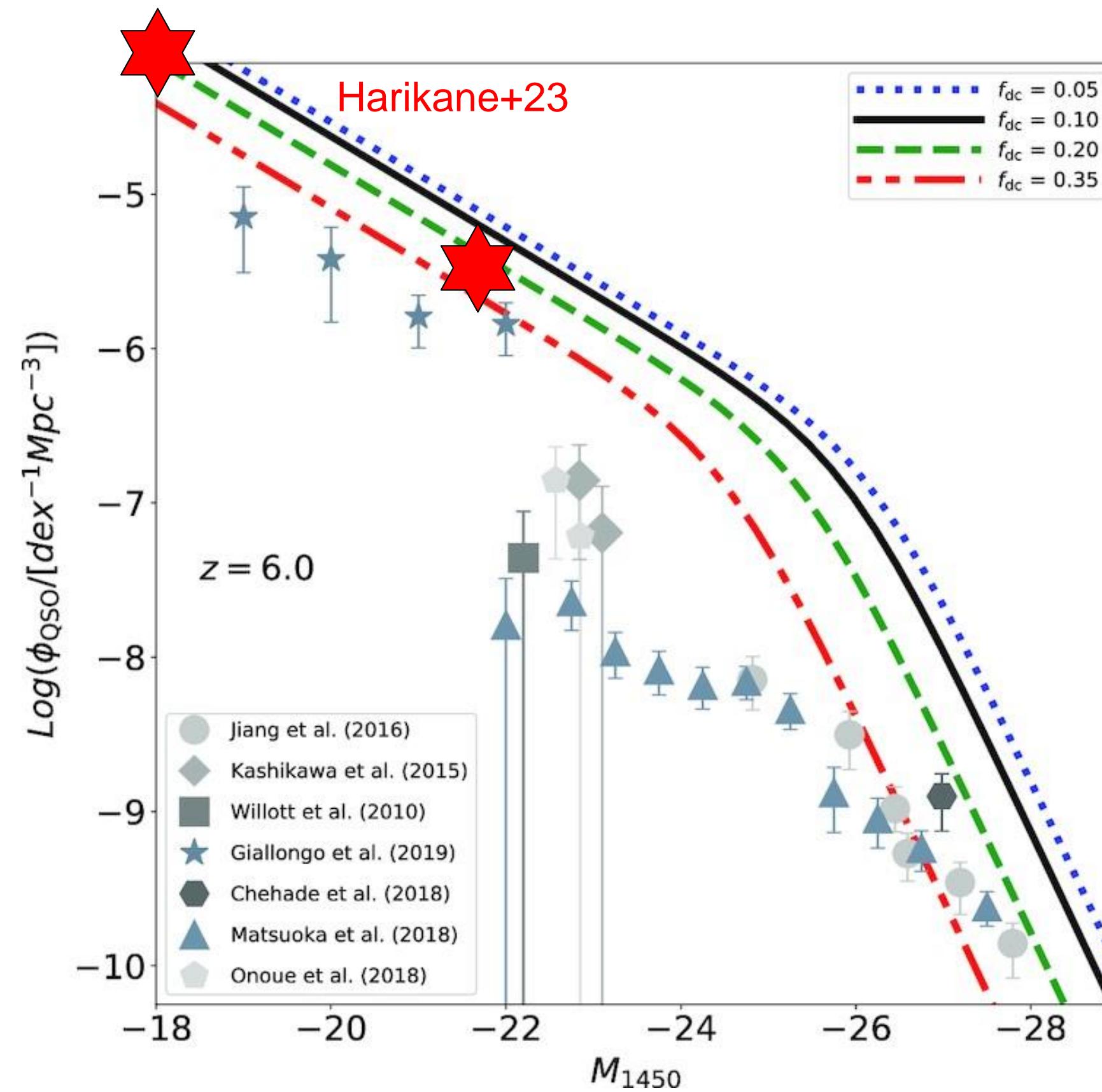
The AGN contribution to the Photoionization rate at $z \sim 5$



Fontanot+2023: HI and Hell neutral fraction



Estimated $z = 6$ AGN LF. The predictions from Fontanot+23 models compared with observations



INTERGALACTIC MEDIUM



The Intergalactic Medium as a particle detector

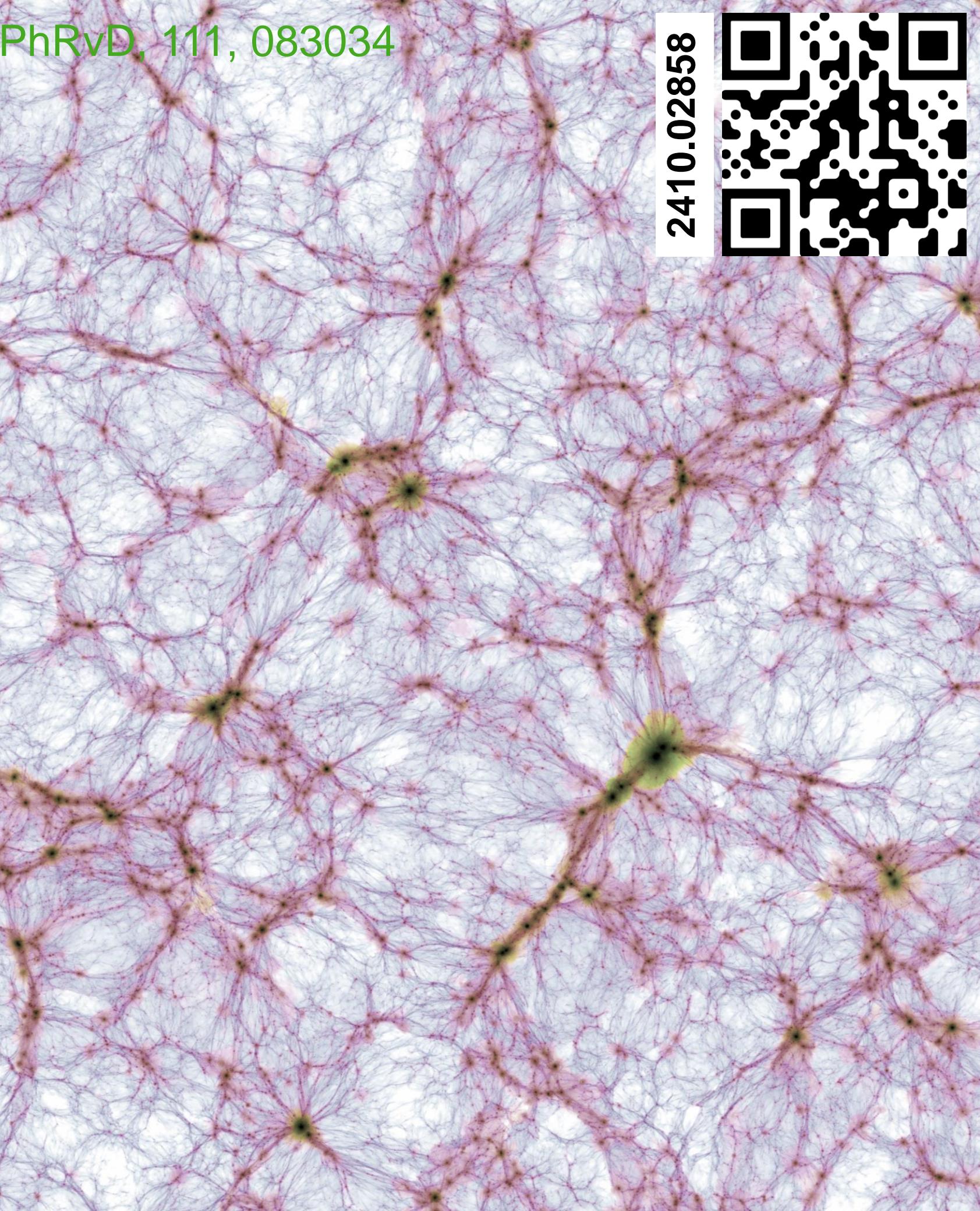
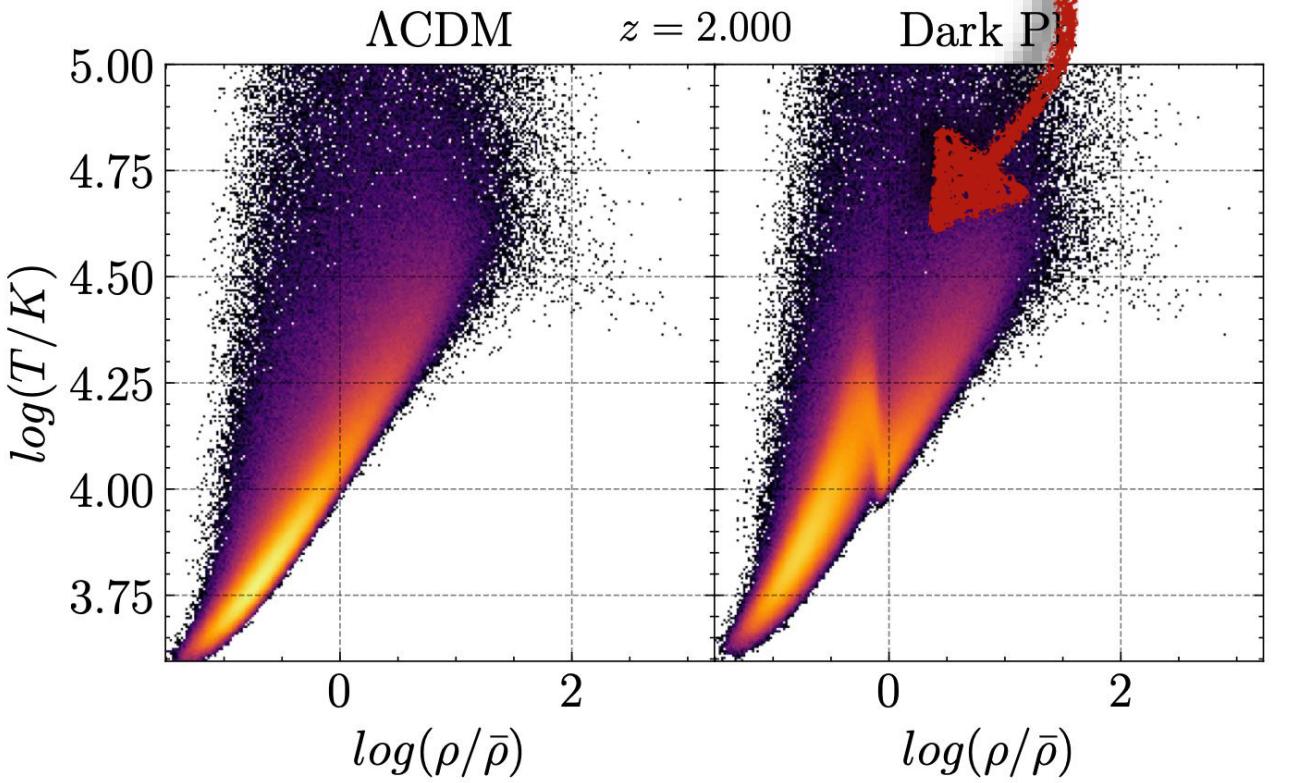
The IGM = all baryons not in galaxies
Mostly ionised hydrogen

Dark Photon

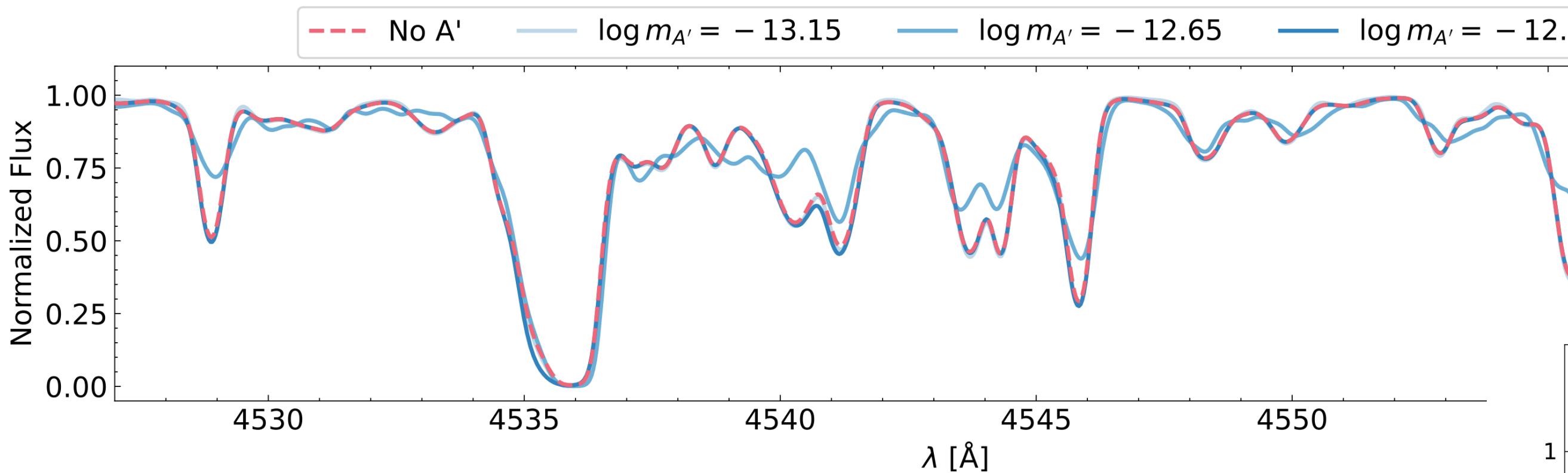
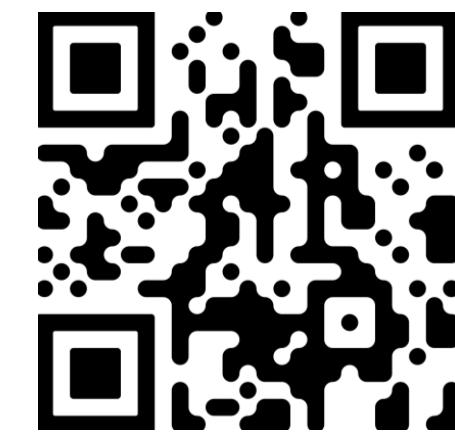
When resonance occurs $m_\gamma^2(n_e) = m_{A'}^2$

Production of γ free-free absorbed by gas

Net energy injection in IGM $E_{A' \rightarrow \gamma}$



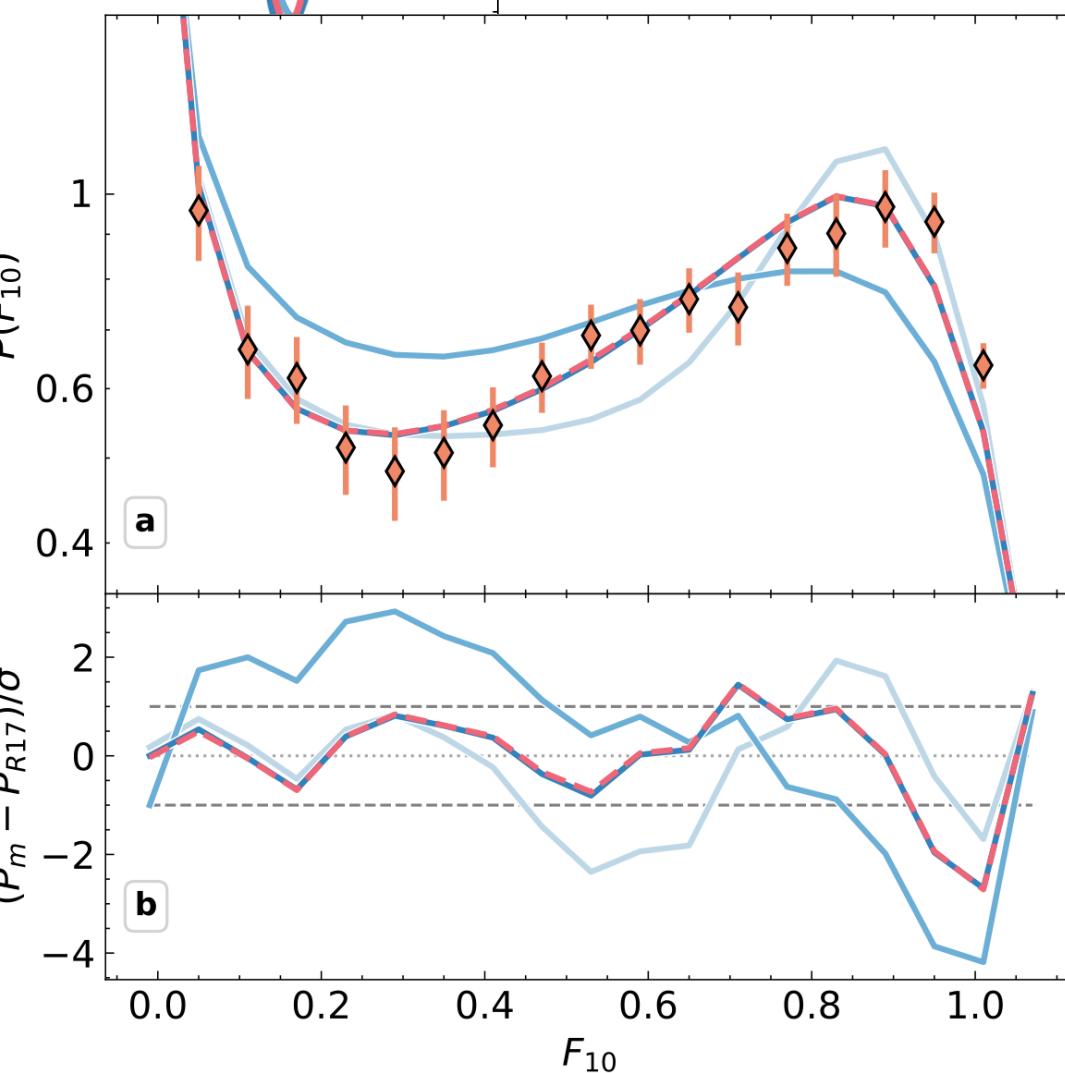
A' in the Lyman Forest



Spectral lines are sensitive to gas temperature

A' deforms the
Transformed and regulated flux PDF

Comparing simulations to high-signal-to-noise
data constraints A' parameters

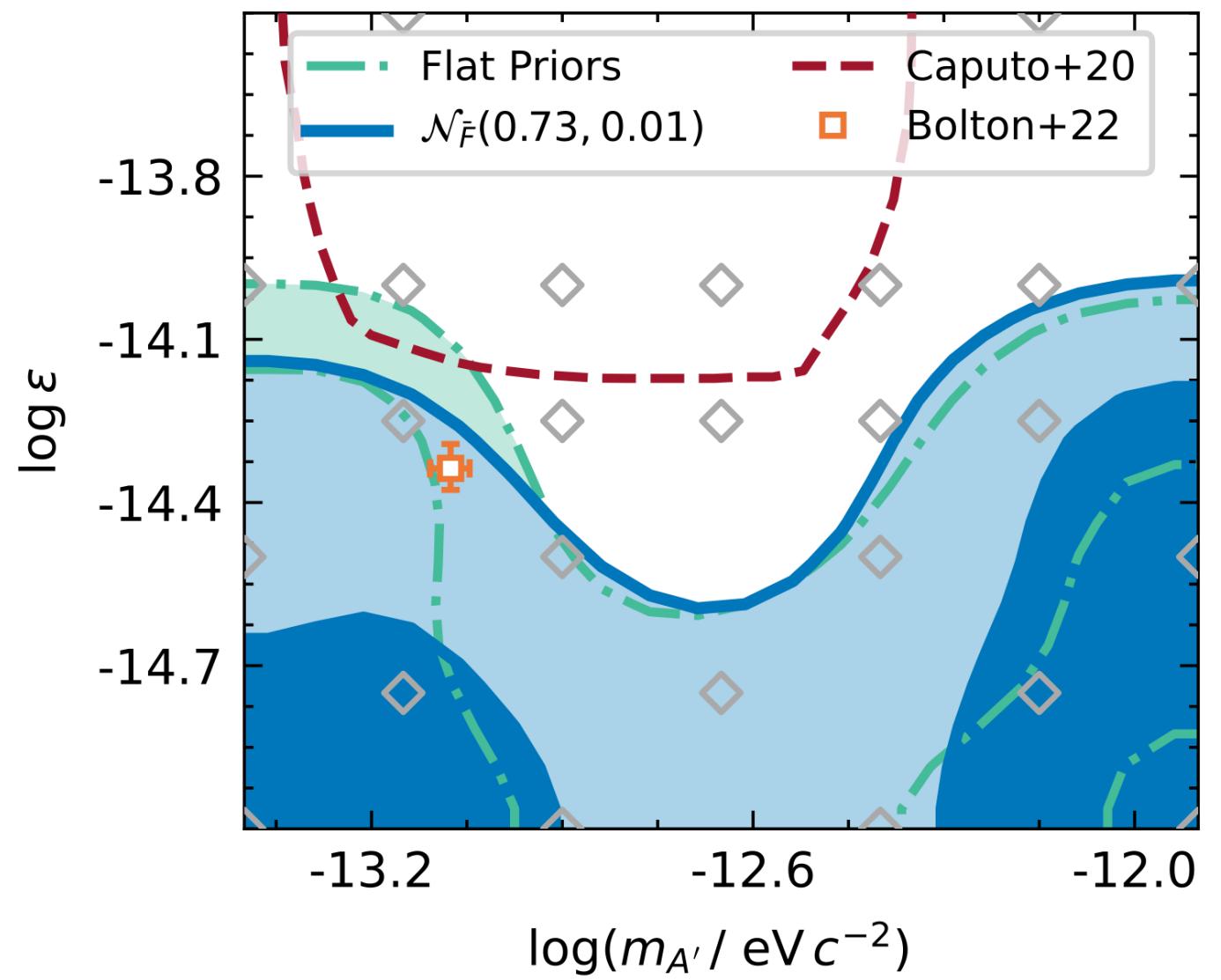


Constraining A'



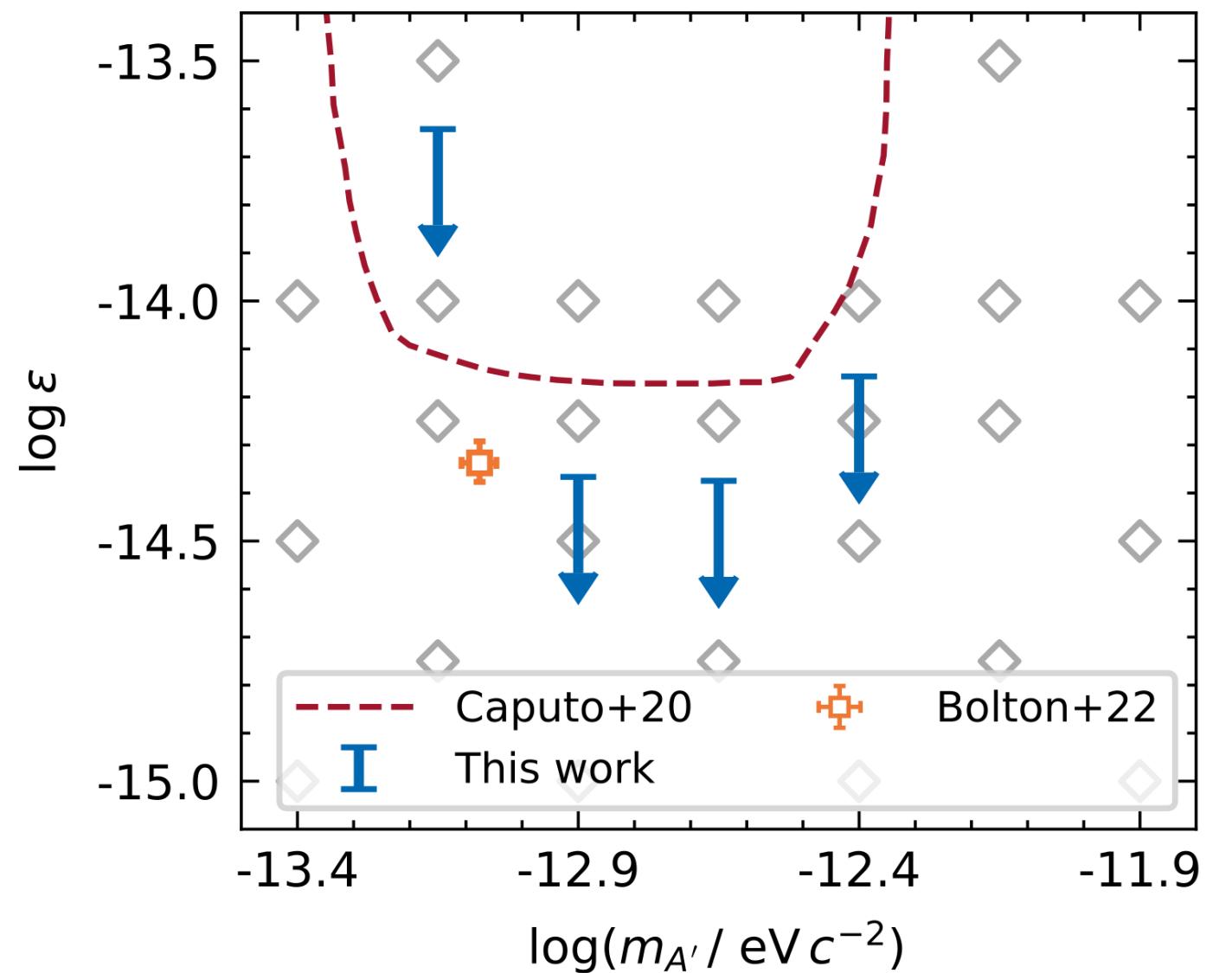
2410.02858

Bayesian — looking for preferred models



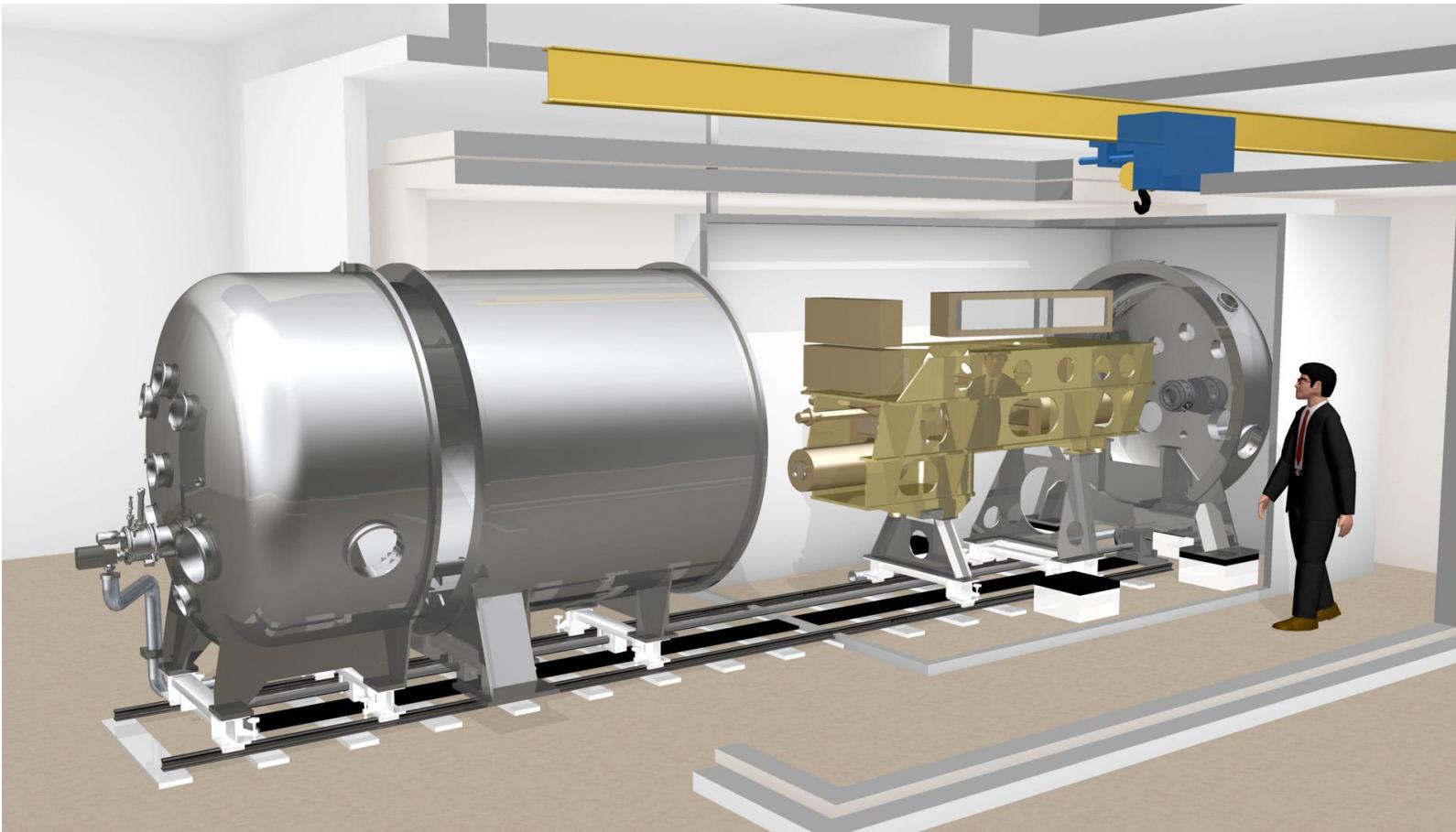
No extra heating required!!

Frequentist — put constraints



New very solid constraints on mass and mixing

ESPRESSO: designed for stability



$$\Delta RV = 1 \text{ m/s}$$

$$\Delta T = 0.01 \text{ K}$$

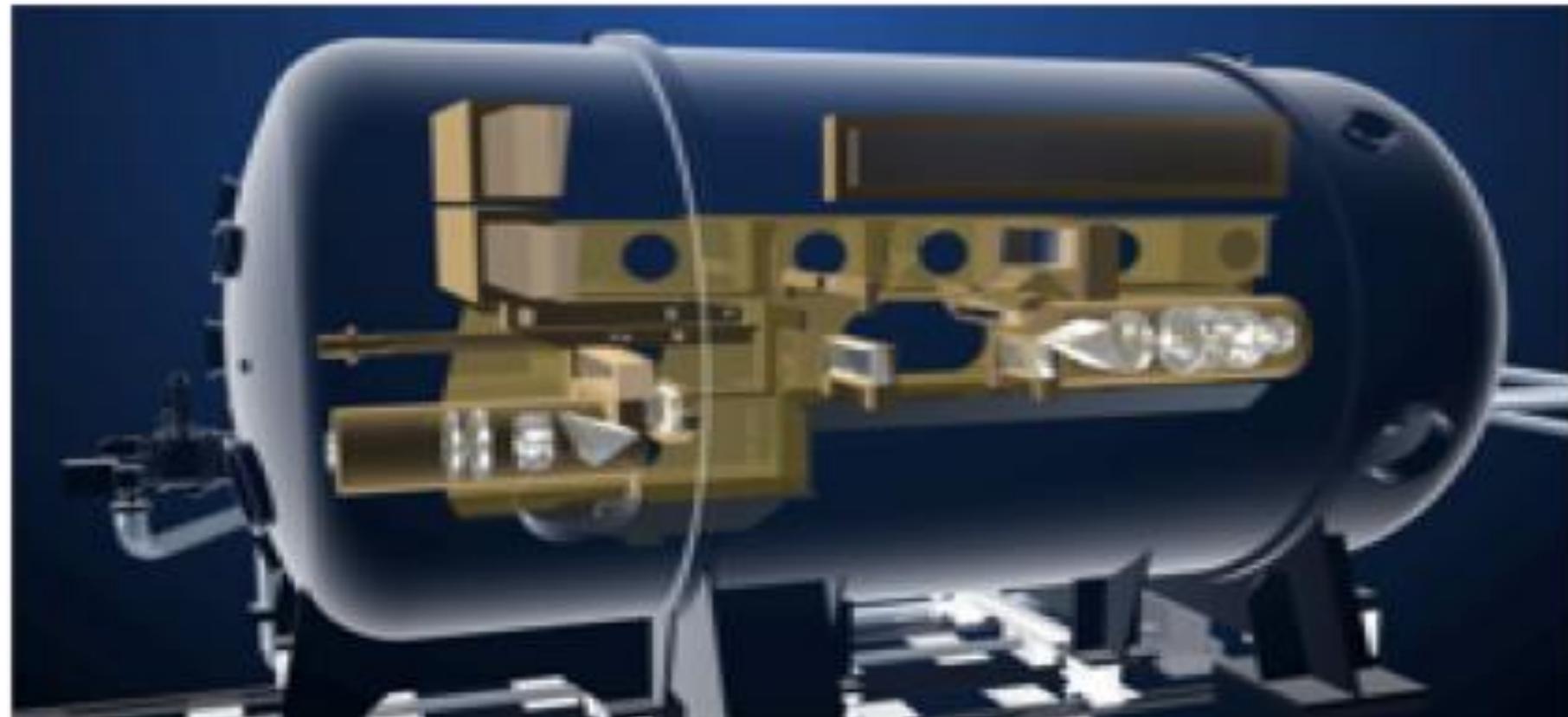
$$\Delta p = 0.01 \text{ mBar}$$

$$\Delta RV = 1 \text{ m/s}$$

$$\Delta \lambda = 0.00001 \text{ \AA}$$

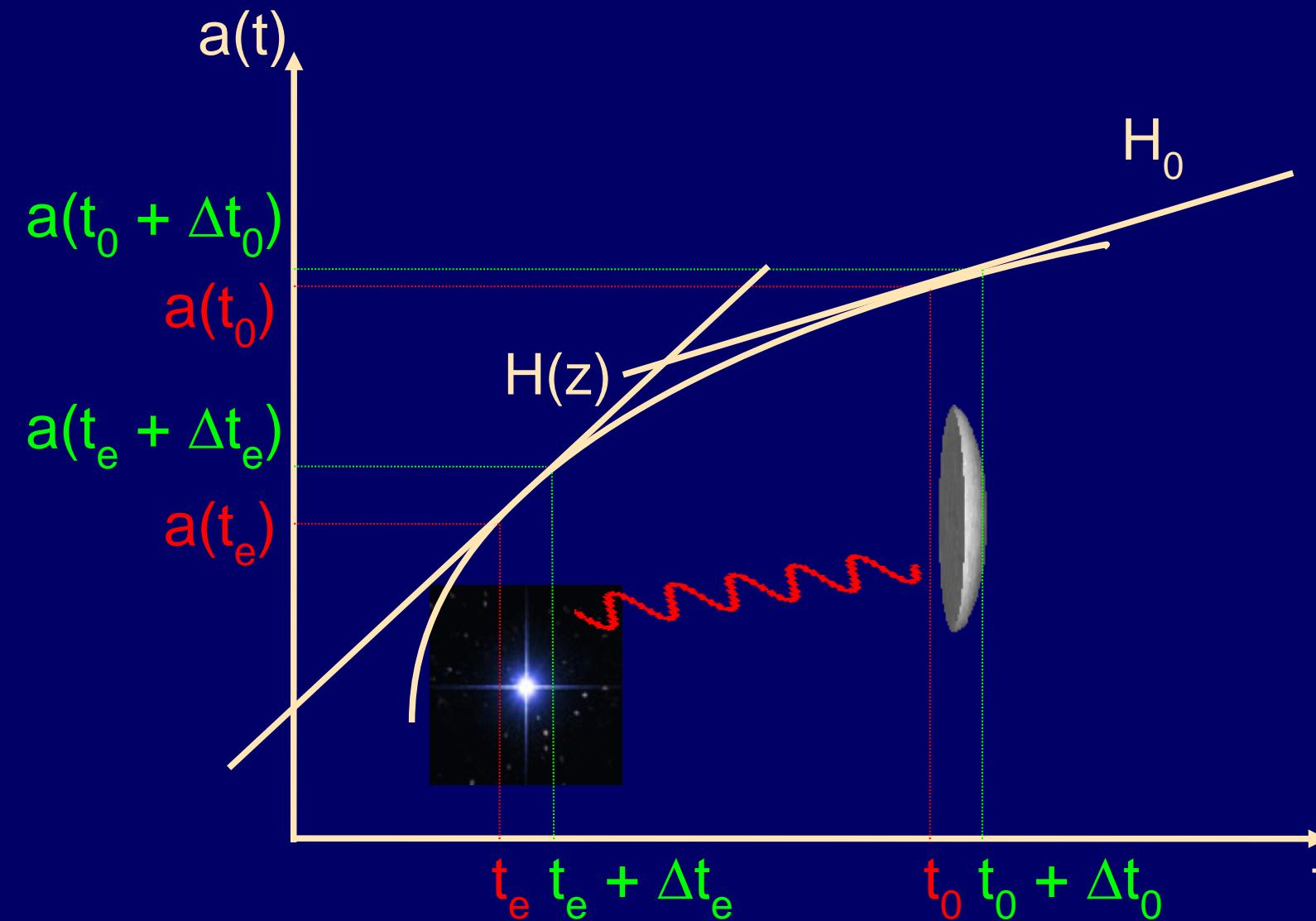
$$15 \text{ nm}$$

$$1/1000 \text{ pixel}$$



The Sandage Test of the Cosmic Redshift Drift

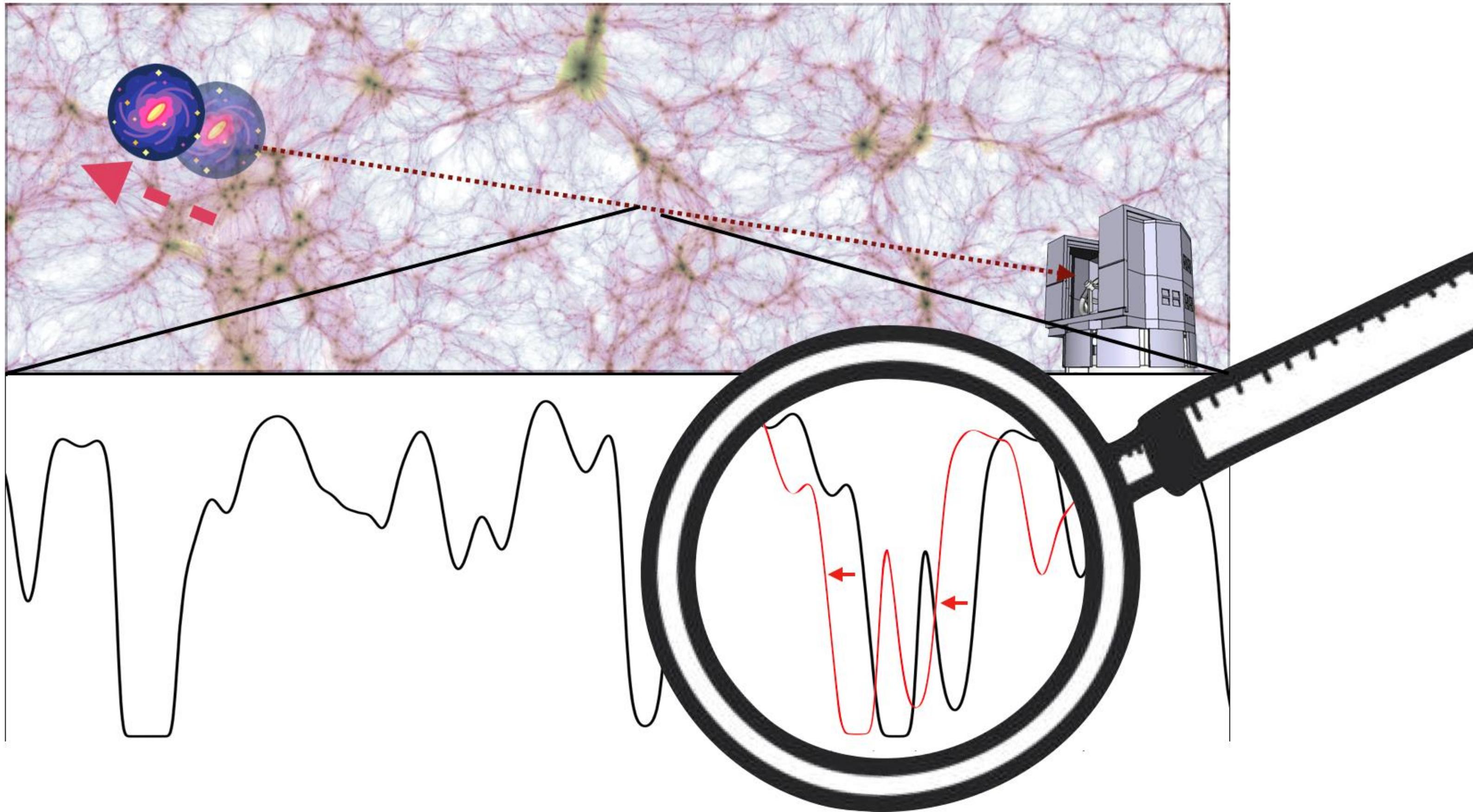
Measuring $H(z)$



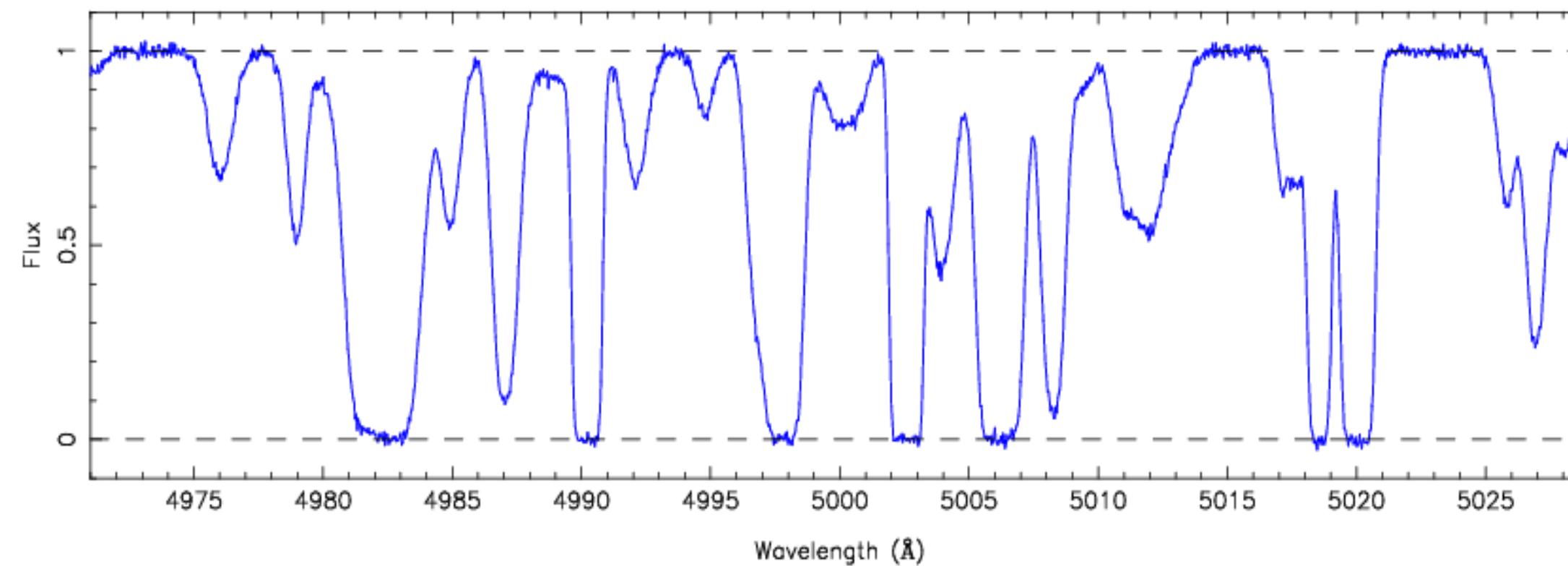
Courtesy J.Liske

$$\frac{z(t_0 + \Delta t_0) - z(t_0)}{\Delta t_0} = \frac{\Delta z}{\Delta t_0} \approx \frac{dz}{dt}_0 = (1+z) H_0 - H(z)$$

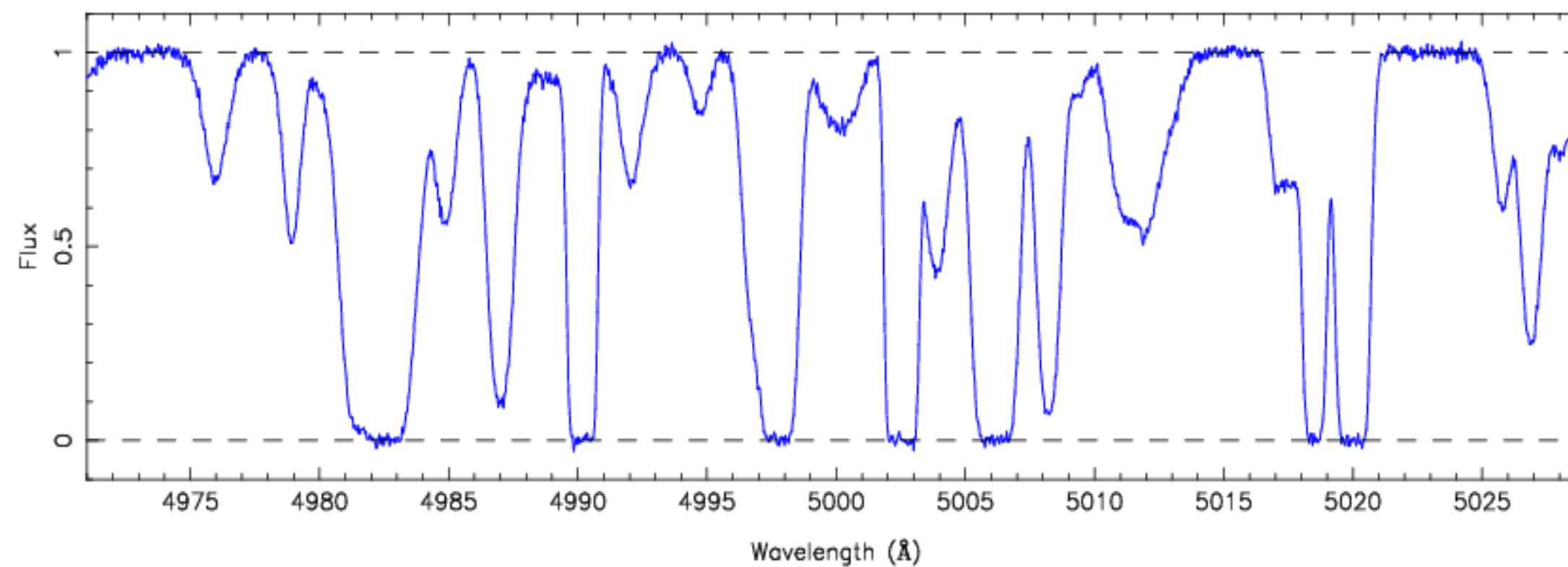
The Lyman Forest - today and ... years after



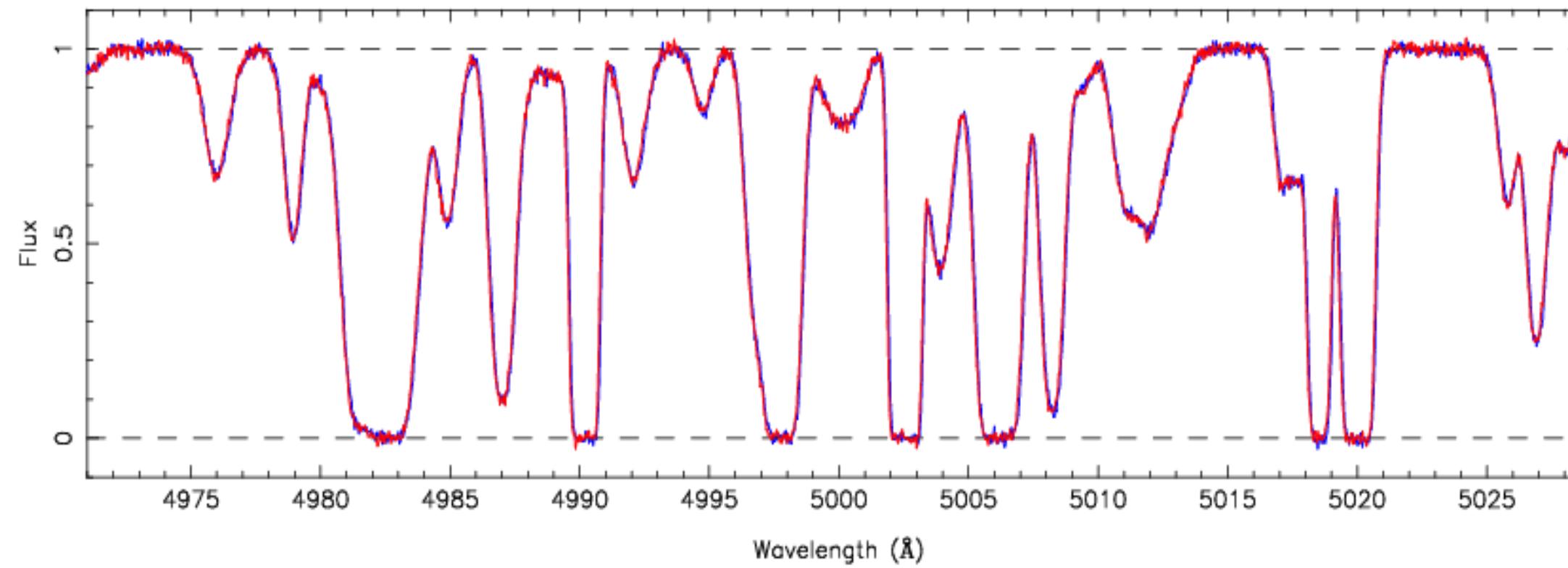
Observing dz/dt in the Ly- α Forest



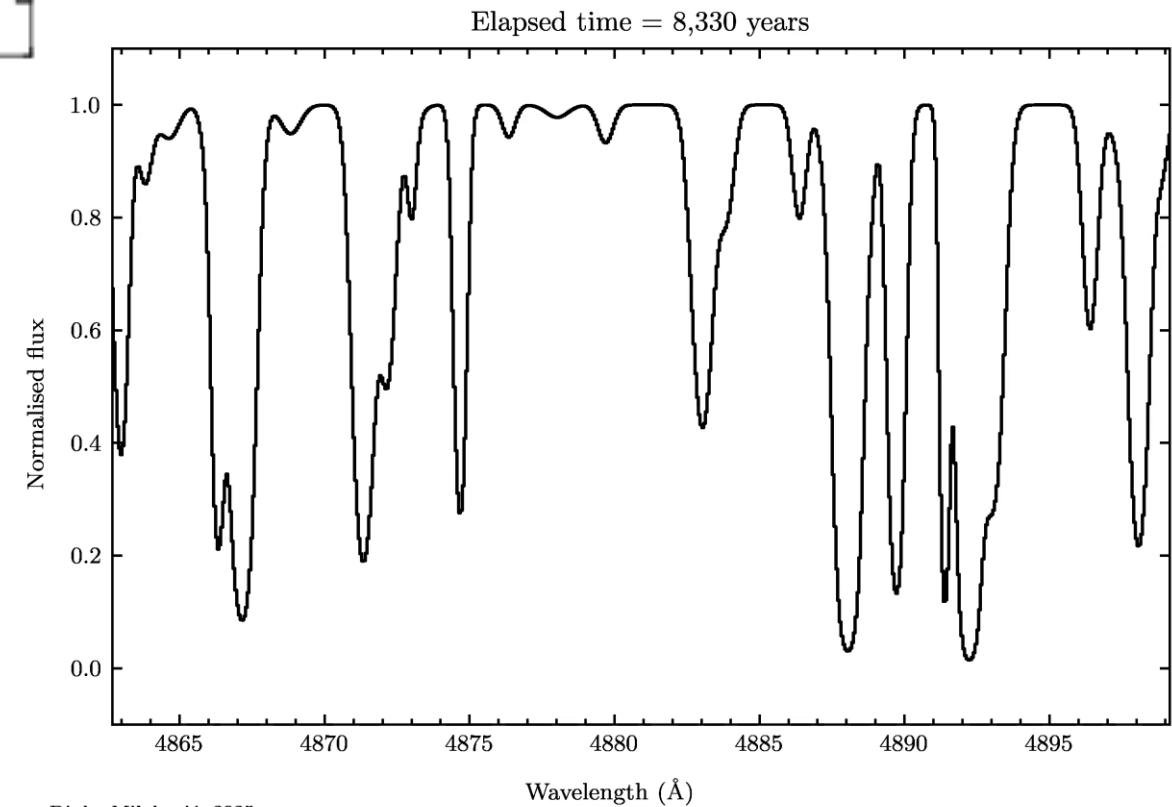
Observing dz/dt in the Ly- α Forest



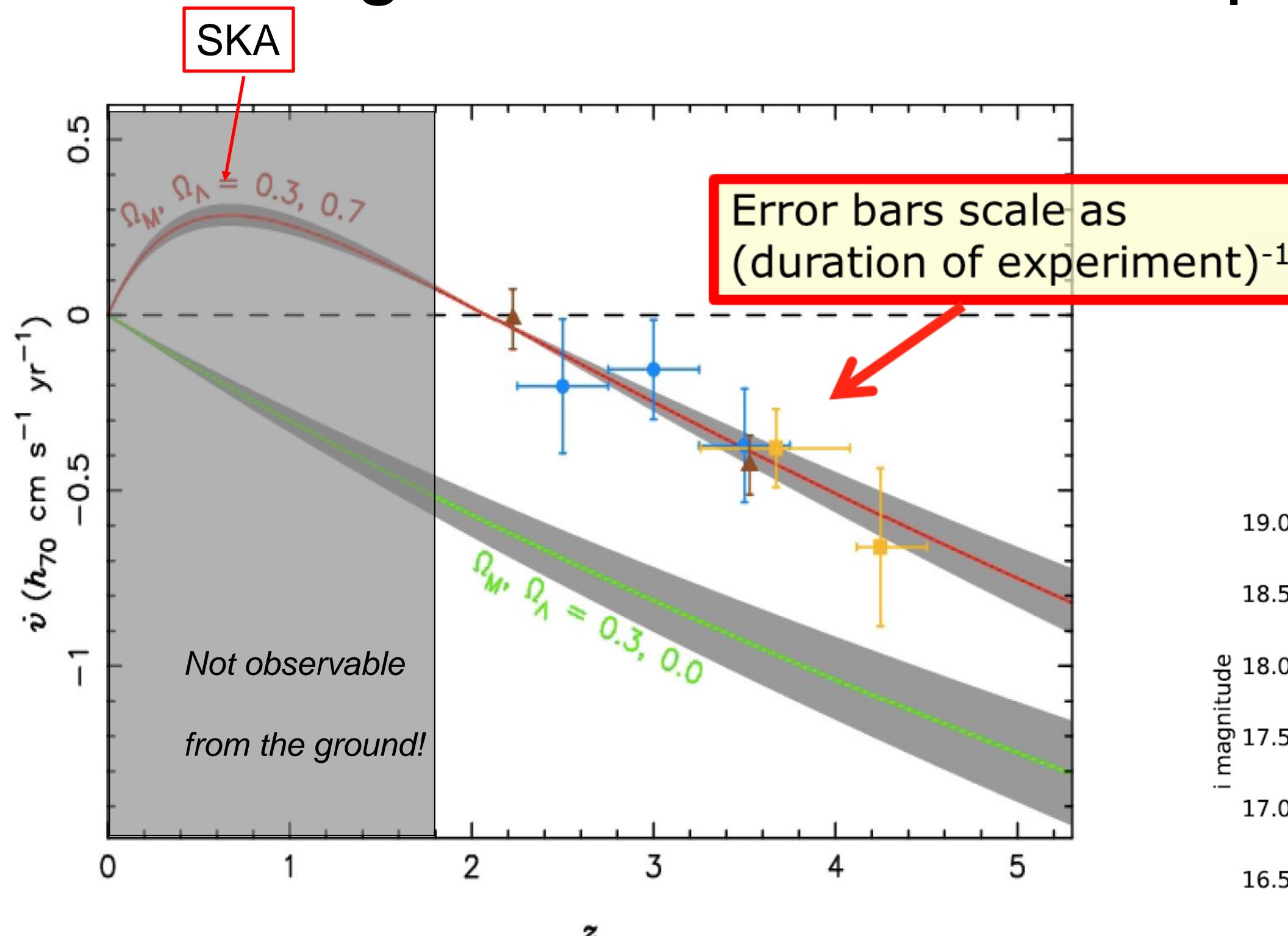
Observing dz/dt in the Ly- α Forest



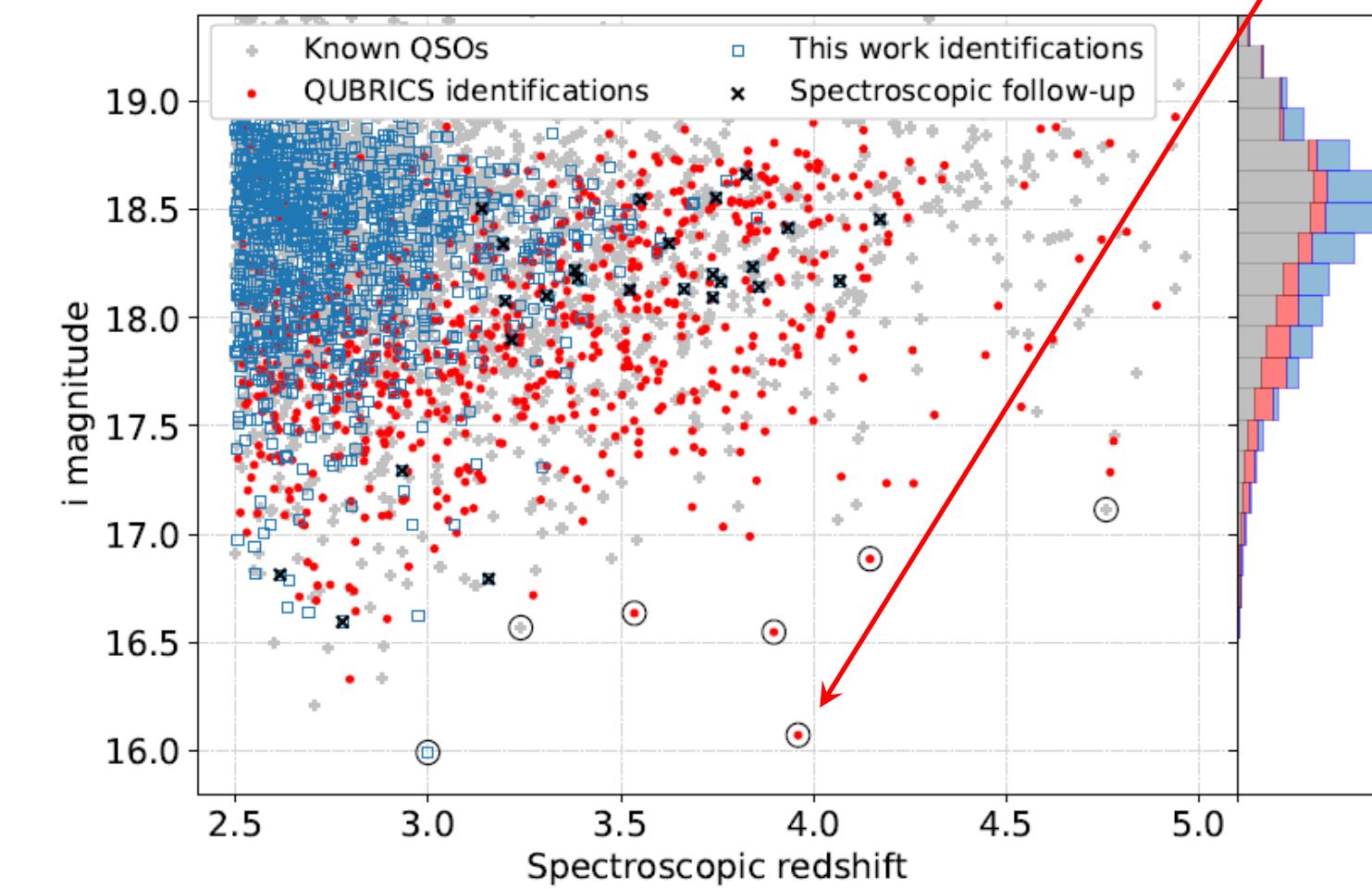
$\Delta t = 10^6$ years!



Sandage Test with a $R_s \sim 10^5$ spectrograph at the E-ELT



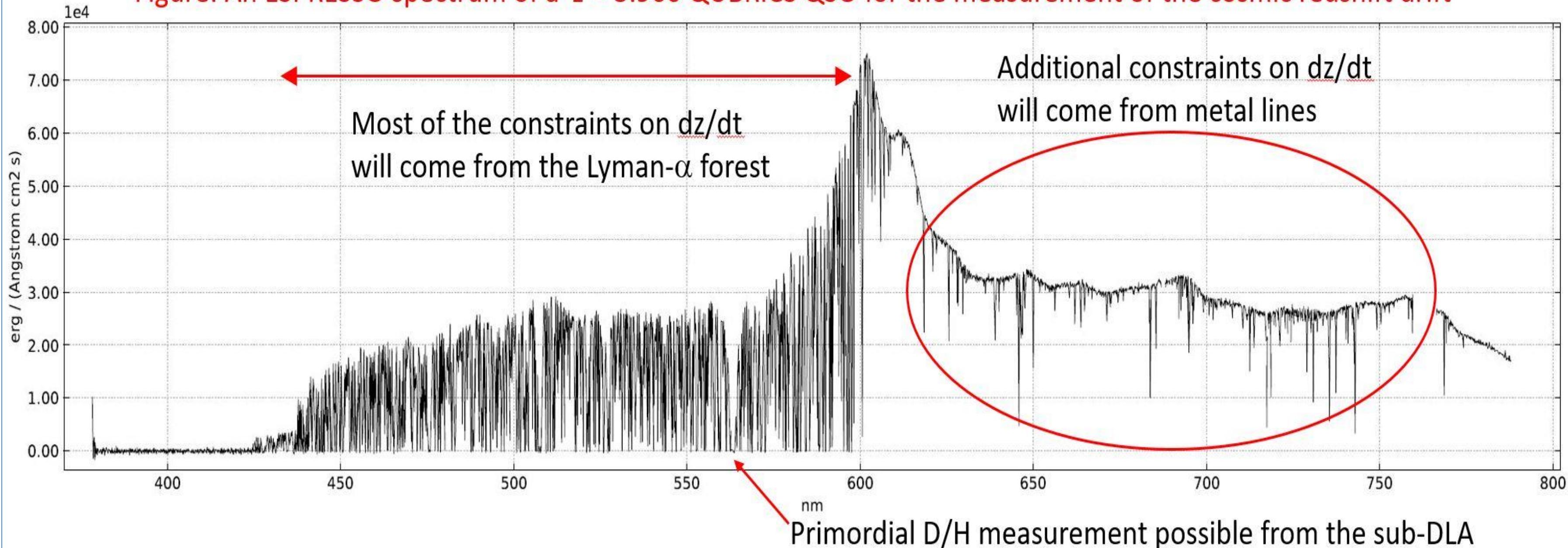
- Different coloured points reflect different targeting strategies
- “**Golden Sample**” of 7 QSOs (Cristiani+2023) from **QUBRICS Survey** (Calderone+2019)



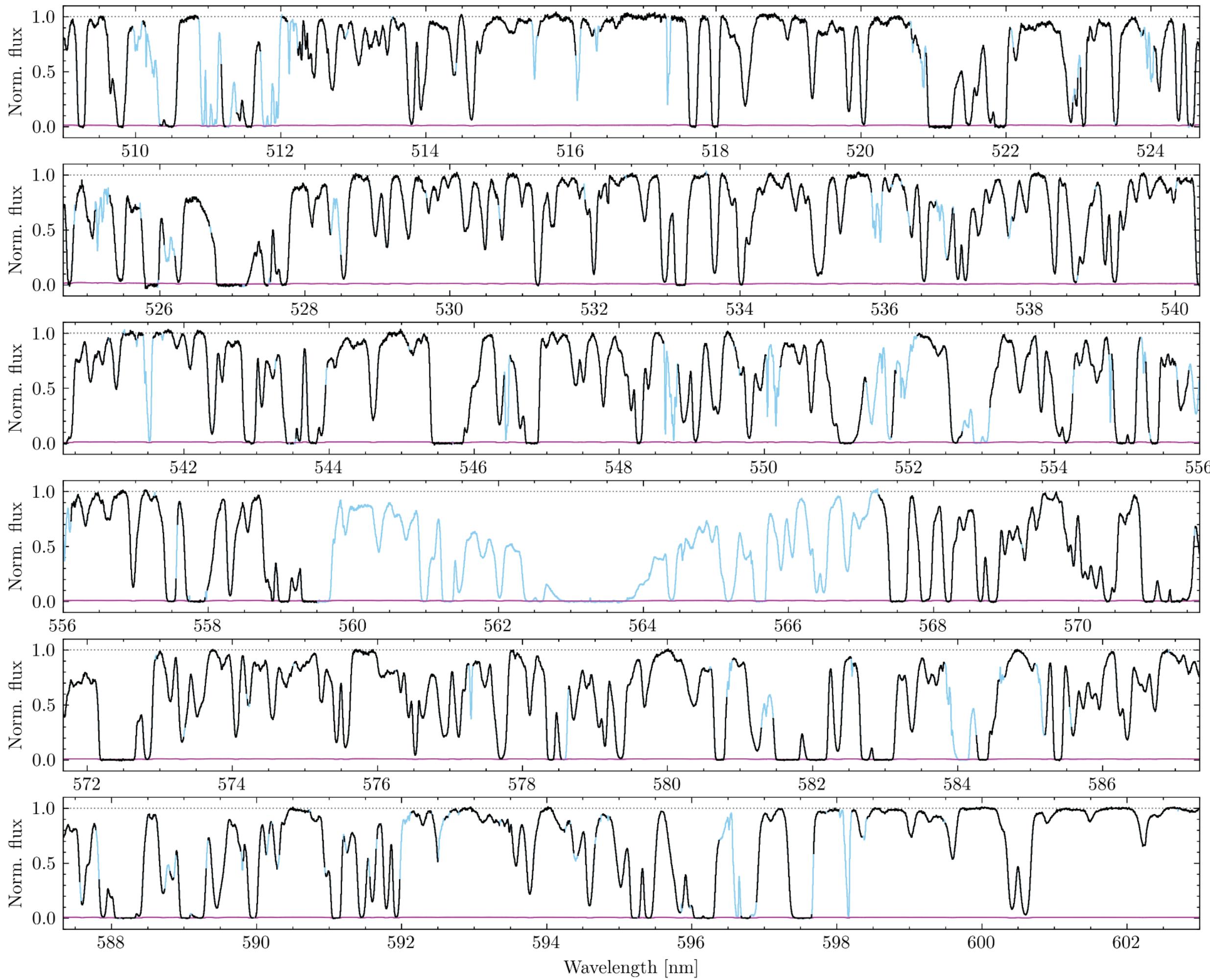
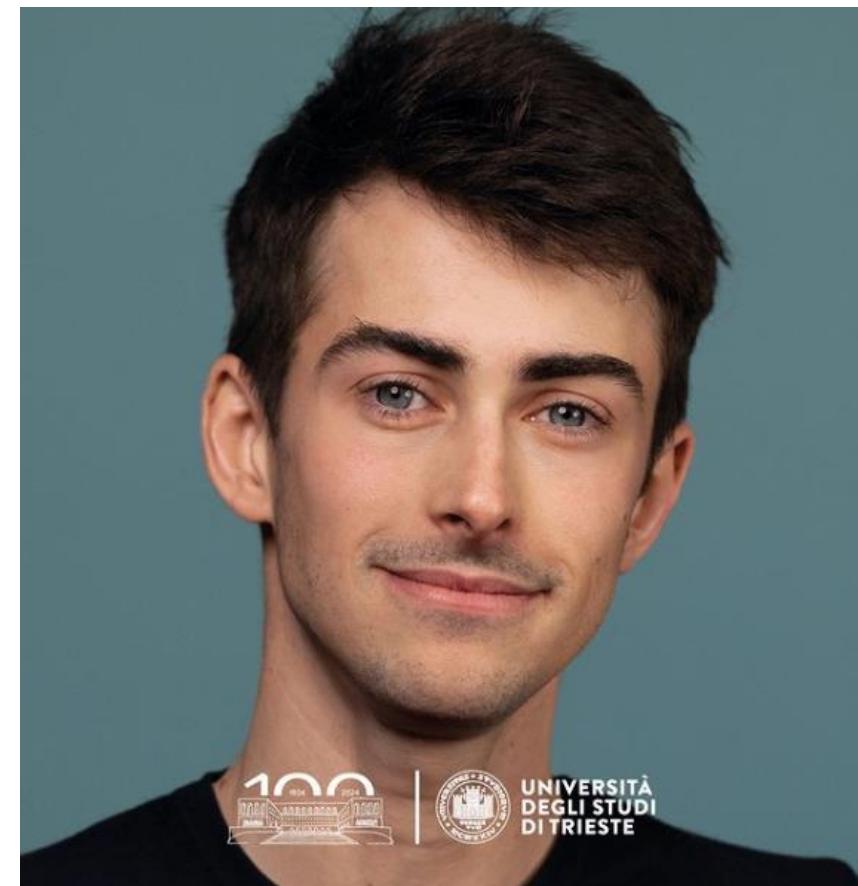
Pasquini et al. 2005, Cristiani et al. 2007, Liske et al. 2008

Sandage test of the cosmic redshift drift

Figure: An ESPRESSO spectrum of a $z = 3.960$ QUBRICS QSO for the measurement of the cosmic redshift drift



SB-2
Trost et al. 2025



$$\Delta v = -1.25^{+4.44}_{-4.46} \text{ m s}^{-1}$$

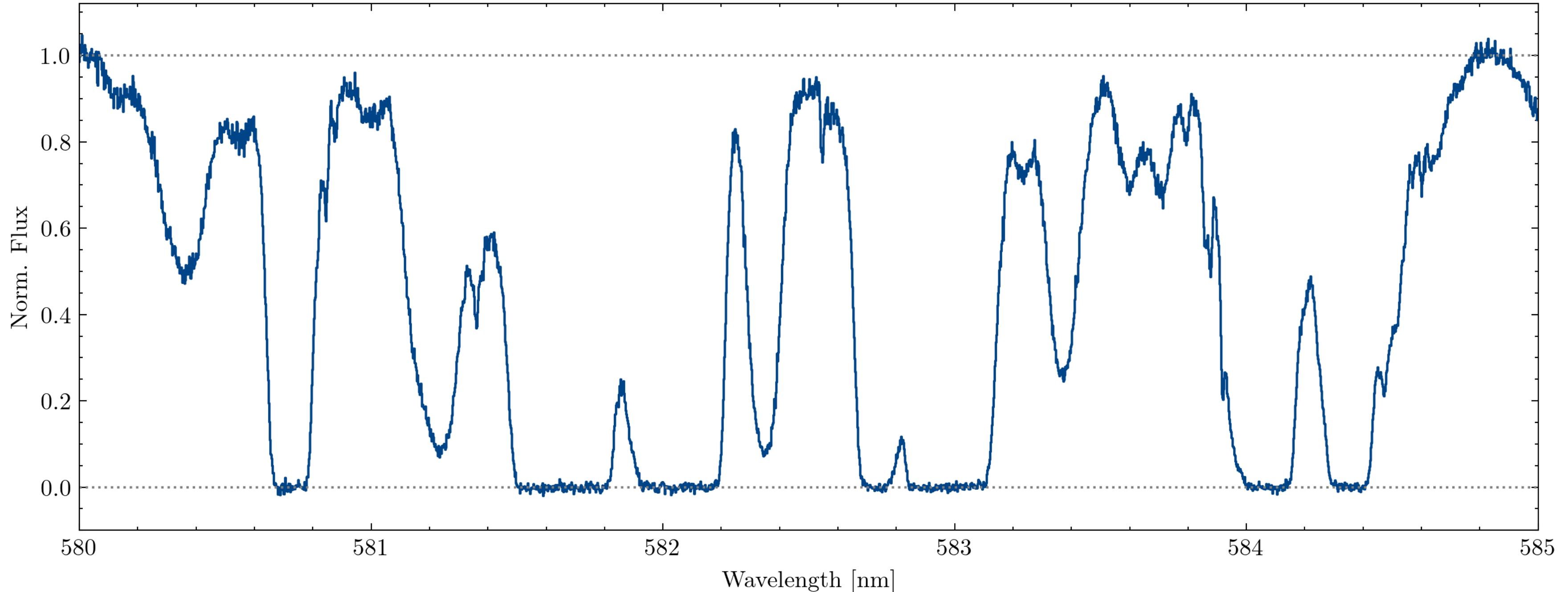
equivalent to

$$\dot{v} = -1.43^{+5.08}_{-5.10} \text{ m s}^{-1} \text{ yr}^{-1}$$

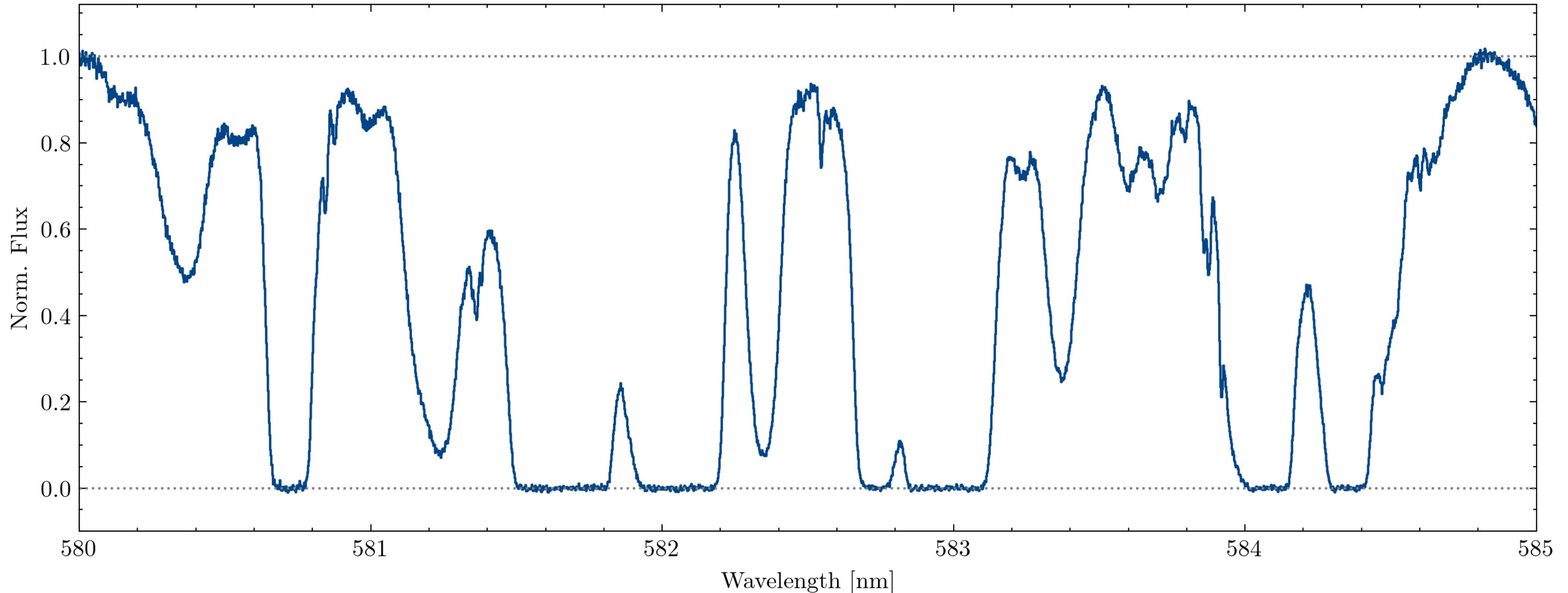
or

$$\dot{z} = -2.19^{+7.75}_{-7.78} \times 10^{-8} \text{ yr}^{-1}$$

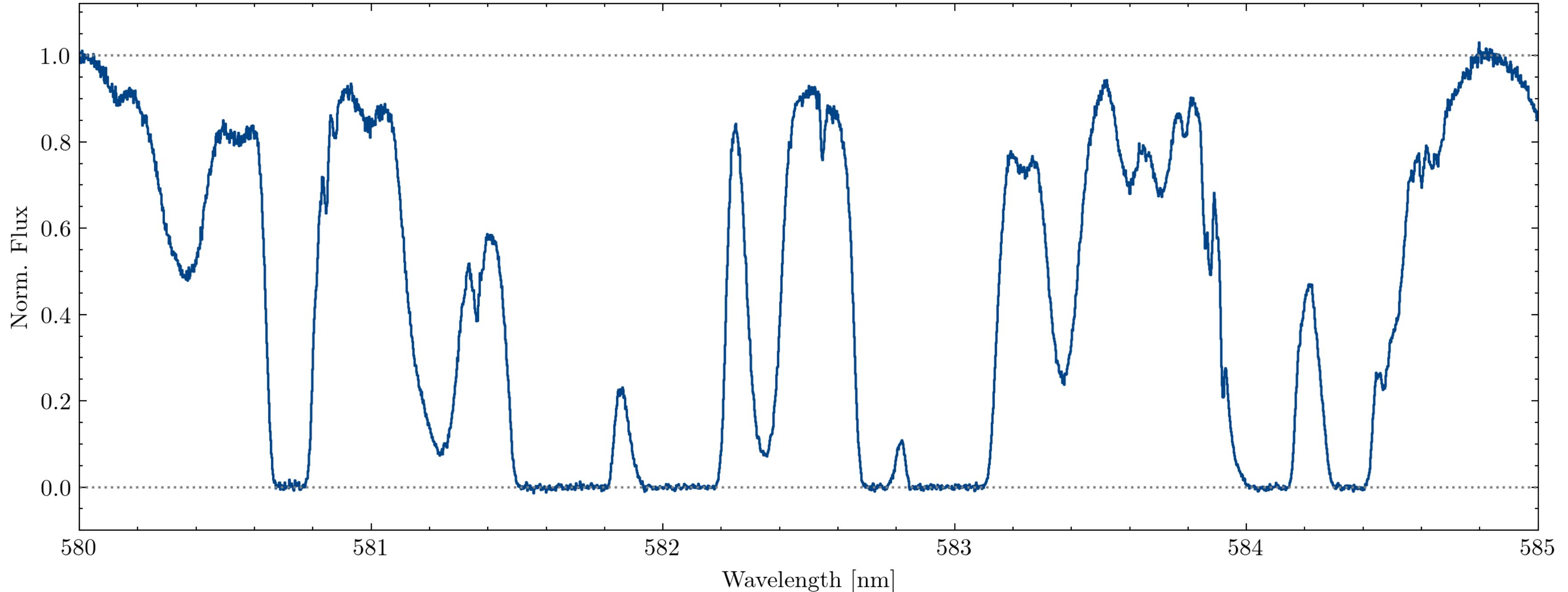
Epoch 1 - 2022



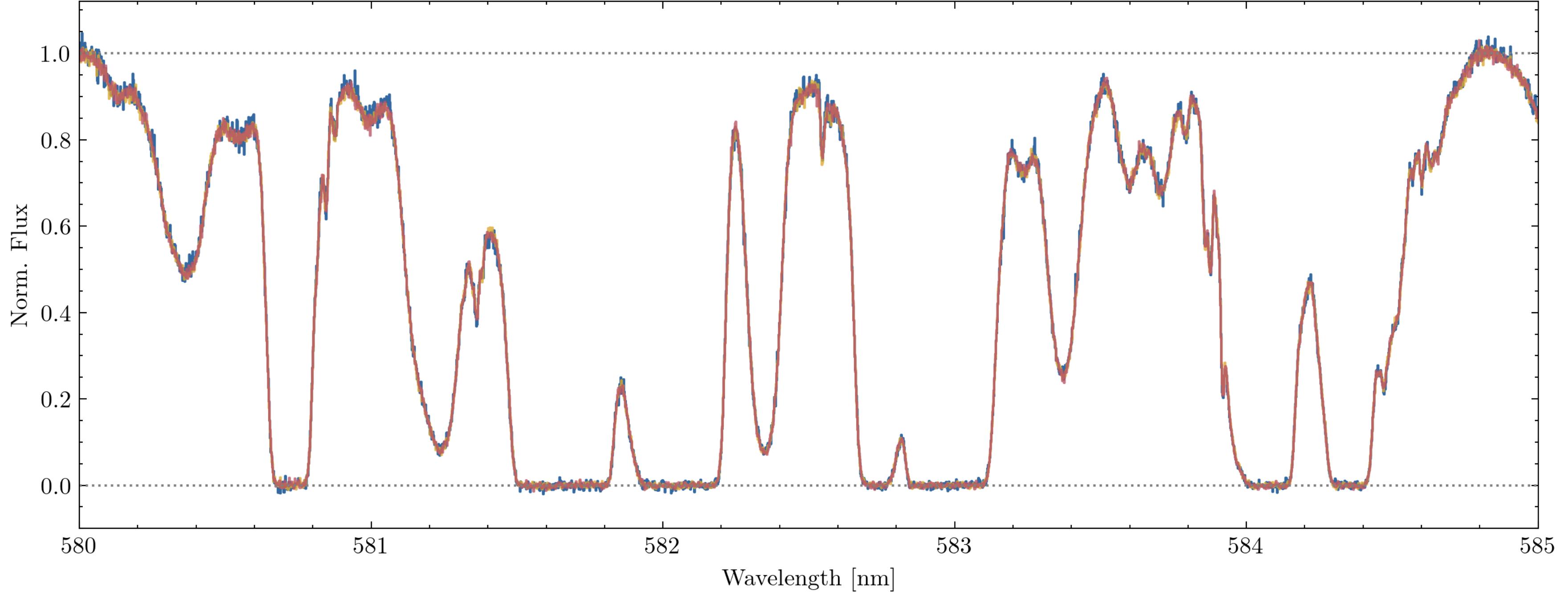
Epoch 2 - 2023



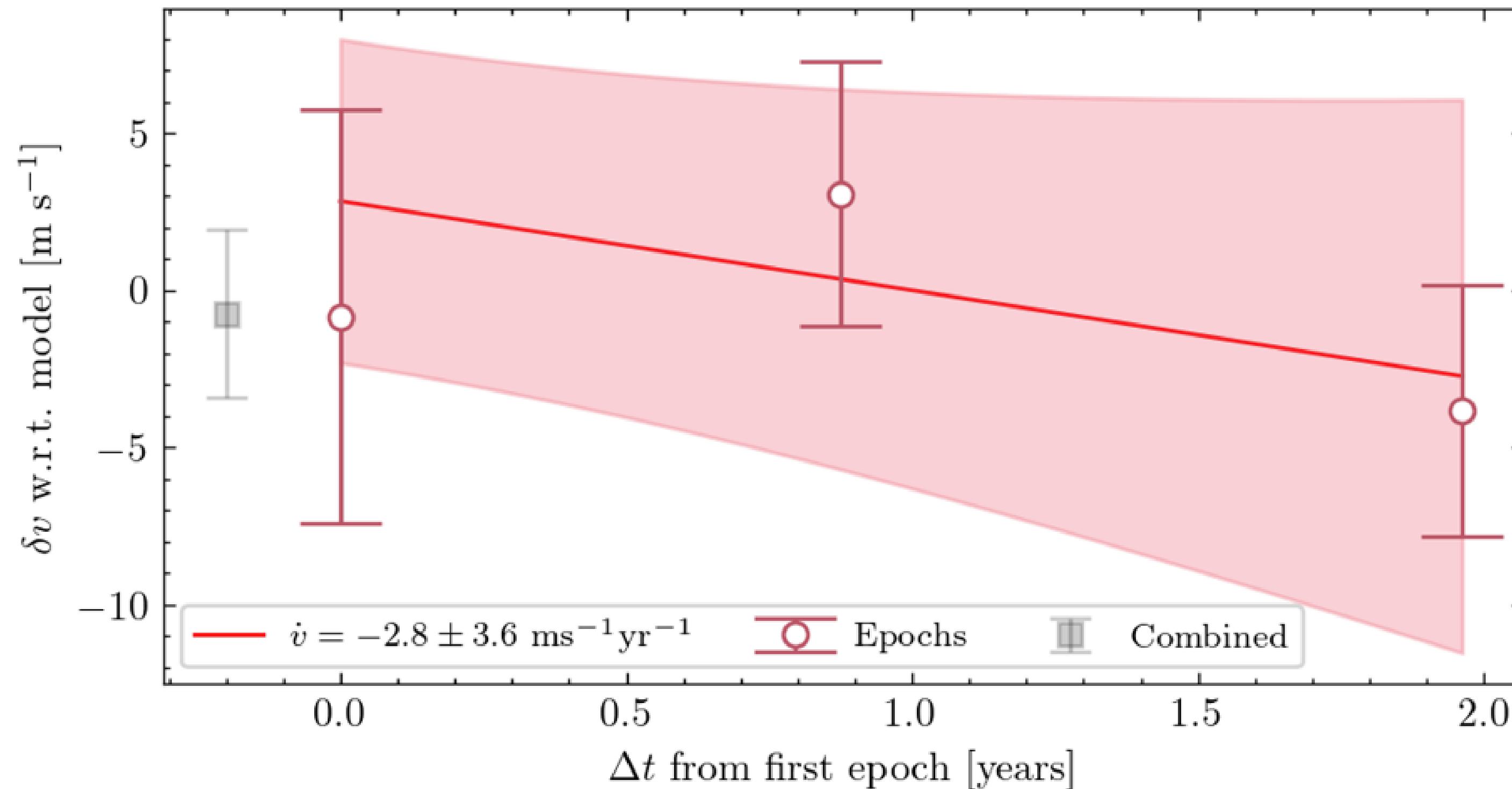
Epoch 3 - 2024



Epoch 1-2-3



SB-2 - Trost et al. in prep

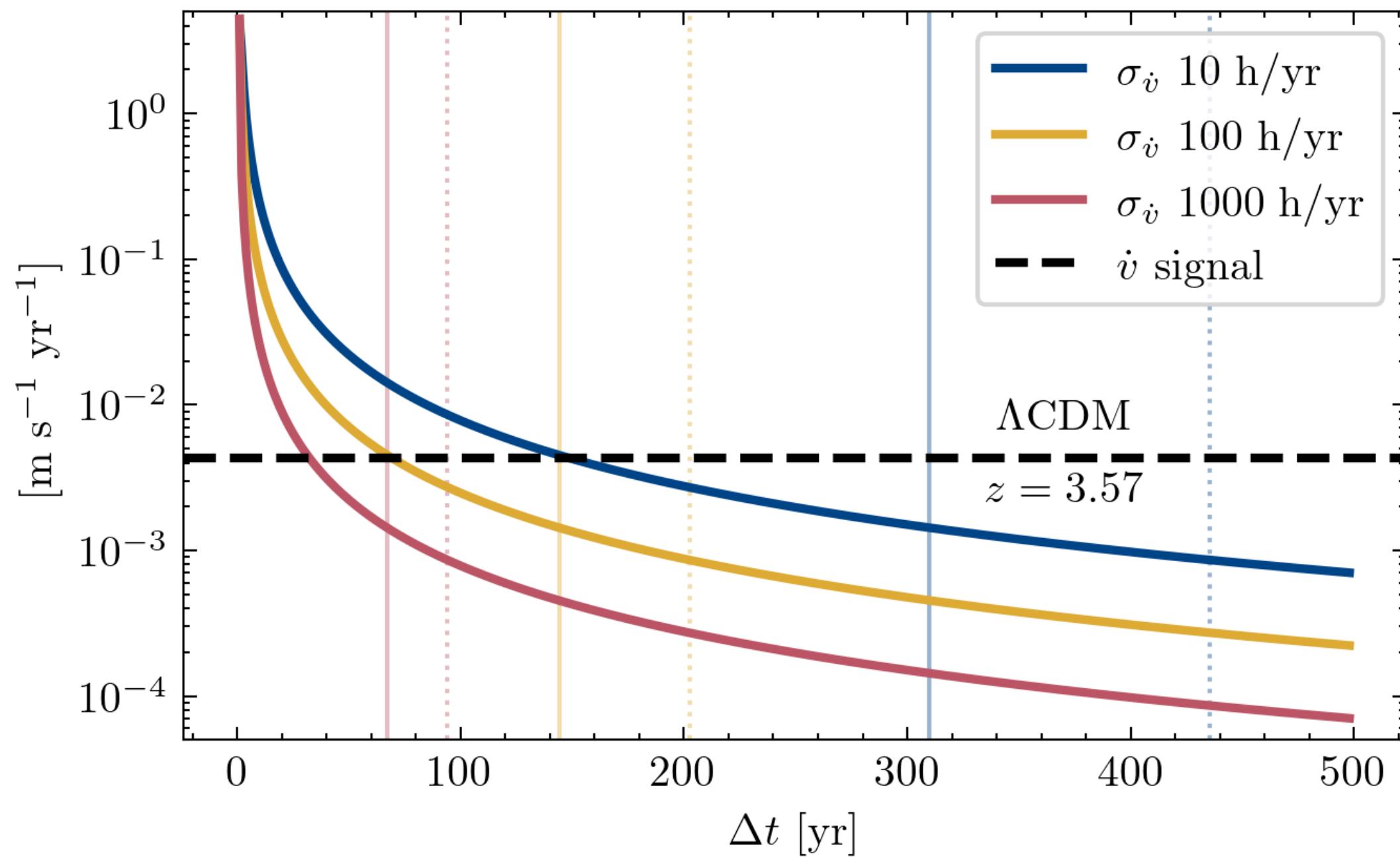


Liske et al. 2008

$$\sigma_v = 2 \left(\frac{S/N}{2370} \right)^{-1} \left(\frac{N_{\text{QSO}}}{30} \right)^{-1/2} \left(\frac{1 + z_{\text{QSO}}}{5} \right)^{-1.7} \text{ cm s}^{-1}$$

$$\frac{S}{N} = 700 \left[\frac{Z_X}{Z_r} 10^{0.4(16-m_X)} \left(\frac{D}{42 \text{ m}} \right)^2 \frac{t_{\text{int}}}{10 \text{ h}} \frac{\epsilon}{0.25} \right]^{1/2}$$

Velocity shift uncertainty reached with ESPRESSO spectra of SB2, assuming three different observational strategies with an integration time of 10 hours per year (blue), 100 hours per year (yellow), and 1000 hours per year (red).



3 σ detection of the redshift drift can be achieved with ESPRESSO(ANDES) after 145 (54) years, assuming 100 hours of observation per year for SB2

$$\Delta t_{3\sigma}^{\text{ANDES}} \approx 54 \left(\frac{T}{100 \text{ h yr}^{-1}} \right)^{-1/3} \left(\frac{\epsilon}{0.1} \right)^{-1/3} \text{ yr}$$

SB-2
Trost et al. 2025



THE
END