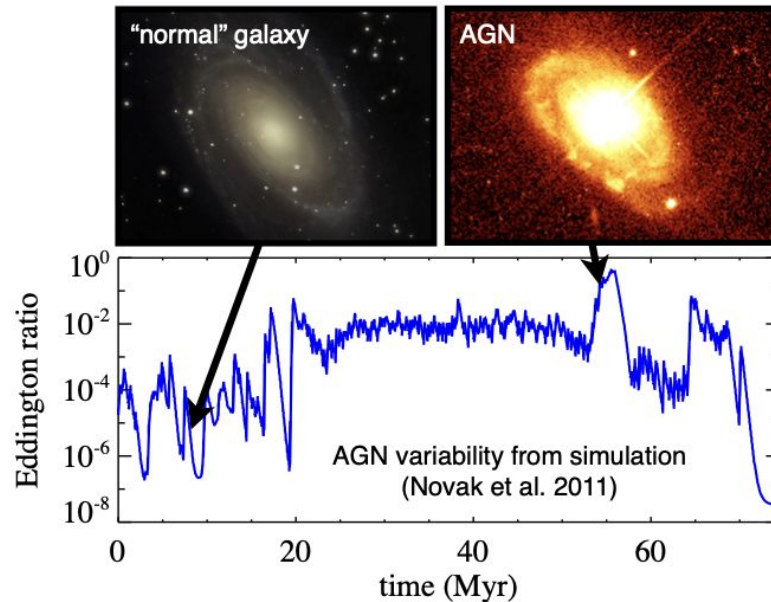


when Active Galactic Nuclei become rare gems

Stéphanie Juneau
NOIRLab

Context

- Most (all) massive galaxies have a supermassive black hole
- Active Galactic Nuclei are found in only a fraction of galaxies
→ BHs are only “active” some of the time (*flicker* on and off)

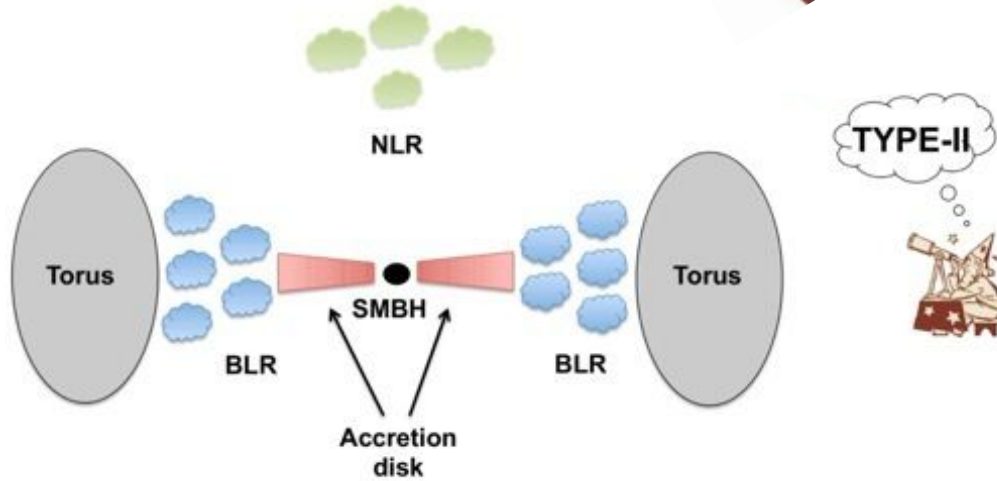


(Hickox et al. 2014)

Context

- Most (all) massive galaxies have a supermassive black hole
- Active Galactic Nuclei are found in only a fraction of galaxies
→ BHs are only “active” some of the time (*flicker* on and off)
- AGN are found in different “flavors”
 - *Active Galactic Nuclei: What’s in a name?* (Padovani+ 2017)
 - Due to orientation effects and/or evolutionary trends or phases

AGN Unified Model



This version: Claudio Ricci

Original versions:
Antonucci (1993); Urry & Padovani (1995)

Also see:
Netzer (2015); Thorne et al. (2022)

Context

- Most (all) massive galaxies have a supermassive black hole
- Active Galactic Nuclei are found in only a fraction of galaxies
→ BHs are only “active” some of the time (*flicker* on and off)
- AGN are found in different “flavors”
 - *Active Galactic Nuclei: What’s in a name?* (Padovani+ 2017)
 - Due to orientation effects and/or evolutionary trends or phases
- Open questions regarding supermassive black holes:
 - **Formation:** black hole seeds
 - **Growth:** accretion and/or mergers
 - **Feedback:** self-regulation + regulation of host galaxies

Approach

- Leverage large optical spectroscopy surveys
 - SDSS: ~1 million galaxies & quasars
 - DESI: ~1 million galaxies & quasars (EDR) and >10 million (DR1; 2025)
- **Part I:** Compare traditional AGN diagnostics to ML model trained on complete spectra (e.g., [Portillo+20](#), [Pat+22](#), [Melchior+23](#), [Liang+23a,23b](#), [Böhm+23](#))
- **Part II:** Investigate DESI target selection: filling in a missing quasar population + finding rare gems ([Juneau+24](#))

Probabilistic Autoencoder (PAE)

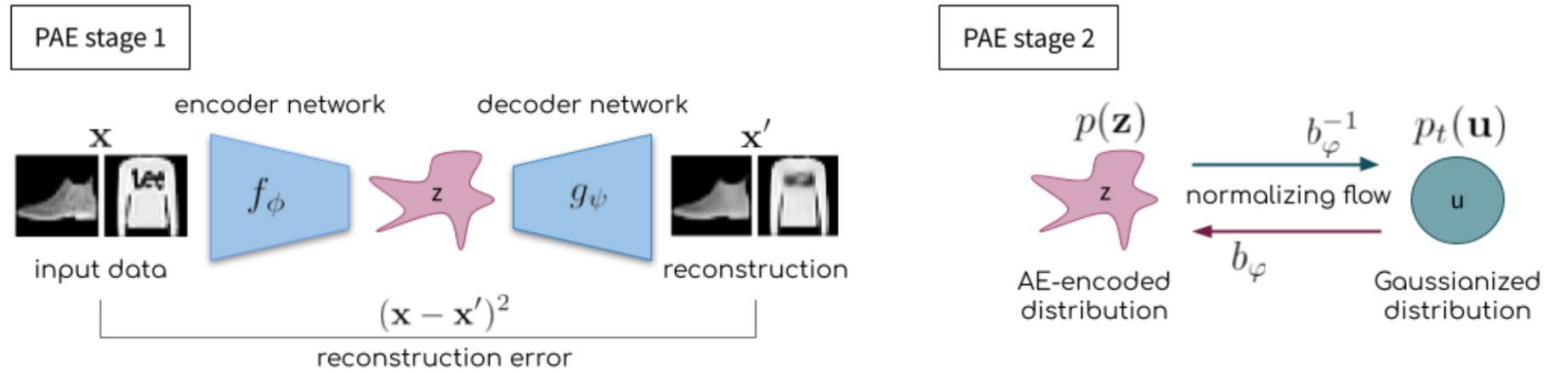


Figure 1. The two-step training of a Probabilistic Autoencoder (PAE) illustrated on an image data set. In the first stage, an autoencoder (AE) consisting of two neural networks is trained to minimize the reconstruction error after compressing (encoding) the data in a lower-dimensional latent space and decompressing (decoding) it back into the high-dimensional data space. In the second stage, a normalizing flow (NF) is trained to learn a bijective mapping from the AE-encoded latent space to a space in which the encoded data follows a Gaussian distribution.

<https://github.com/VMBoehm/PytorchPAE>

(Böhm & Seljak 2022; Böhm, Kim & Juneau 2023; also see a similar PAE approach by Melchior et al. 2023)

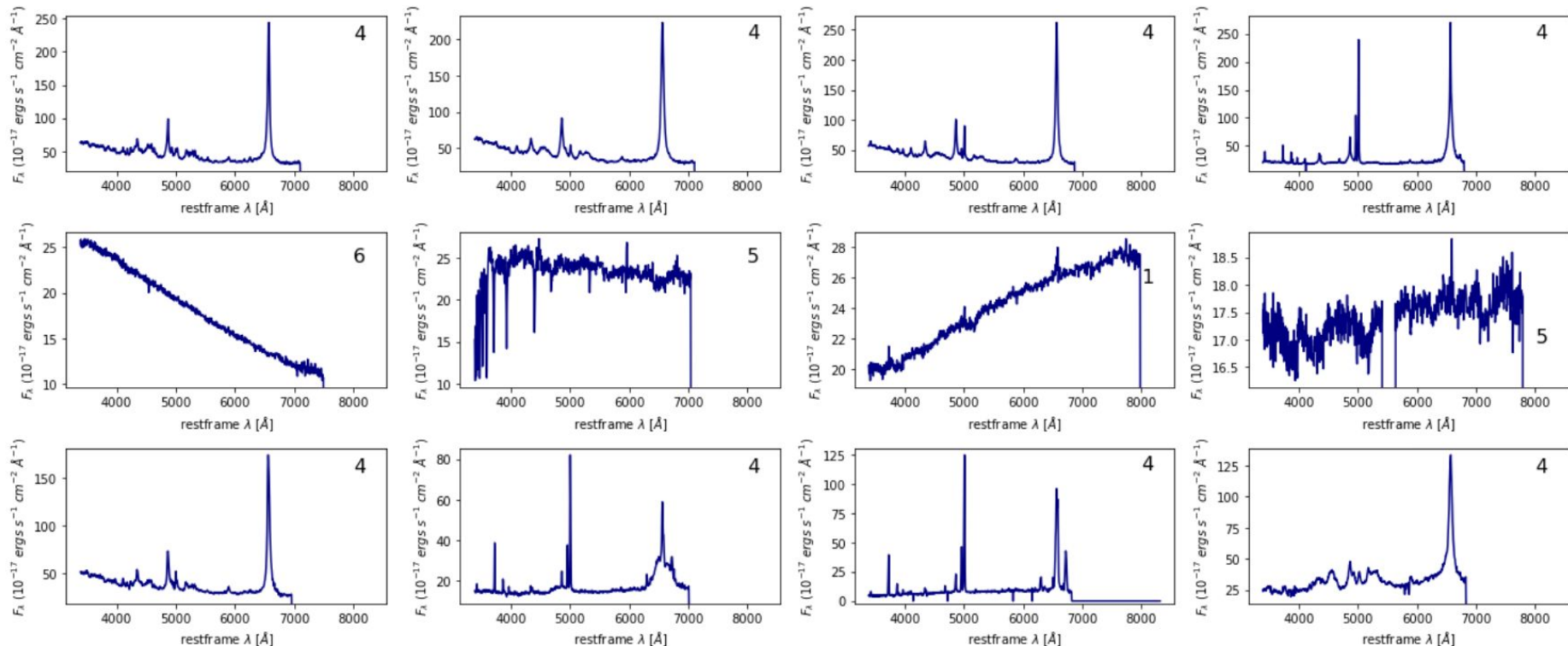
Probabilistic Autoencoder (PAE)

- Strategies to account for certain classes (e.g., QSO) being more rare than others (e.g., GALAXY)
 - Evaluate conditional probability given a label (class+subclass)
- Examples without (e.g., Pat+22, Liang+23a) and with (Böhm+23) conditional density estimation

<https://github.com/VMBoehm/PytorchPAE>

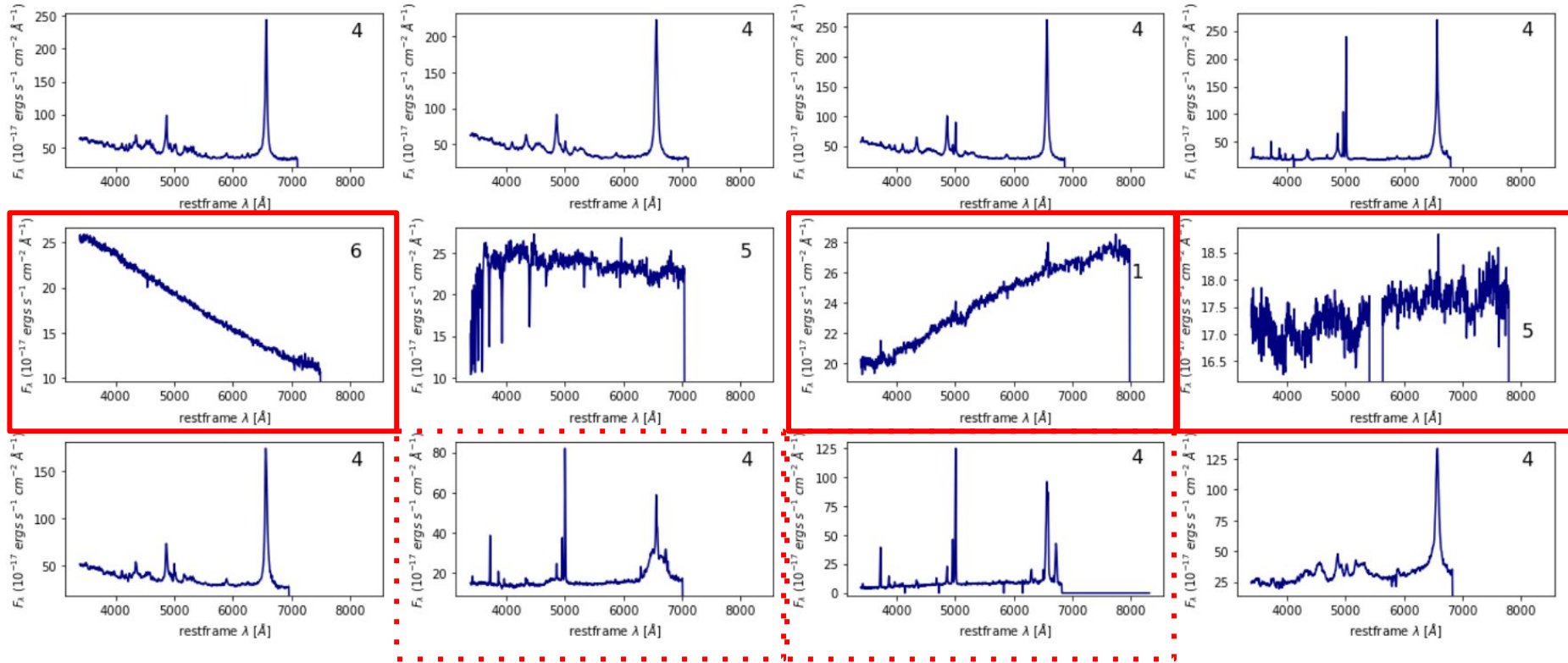
(Böhm & Seljak 2022; Böhm, Kim & Juneau 2023)

Example outlier SDSS spectra (non-conditional on labels)



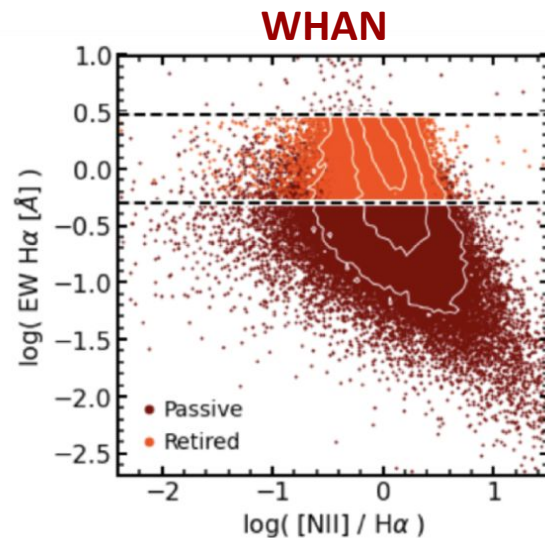
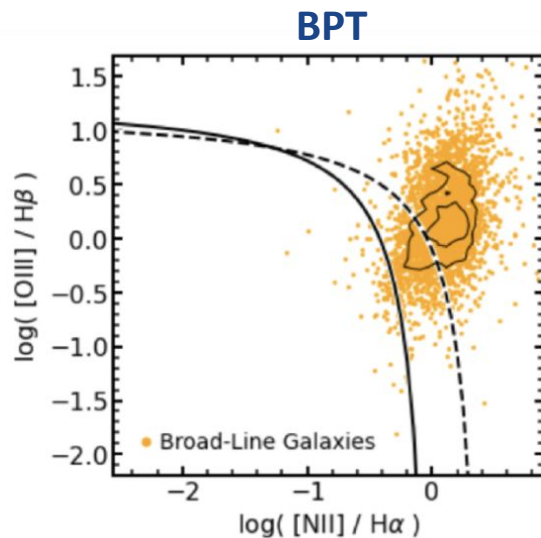
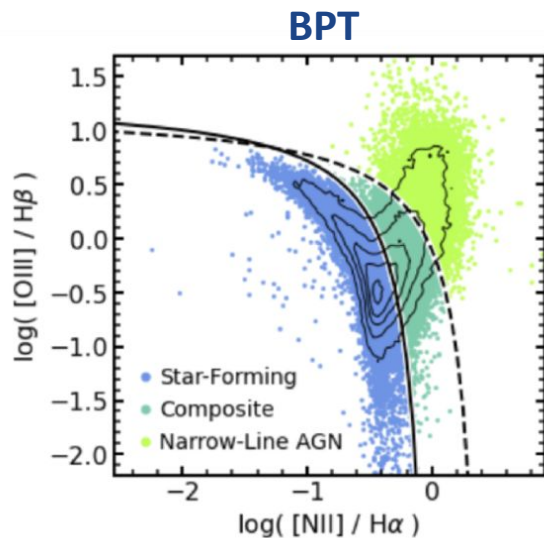
(Pat, Juneau, Böhm et al., 2022)

Example outlier SDSS spectra (non-conditional on labels)



(Pat, Juneau, Böhm et al., 2022)

Classification combining BPT & WHAN diagrams



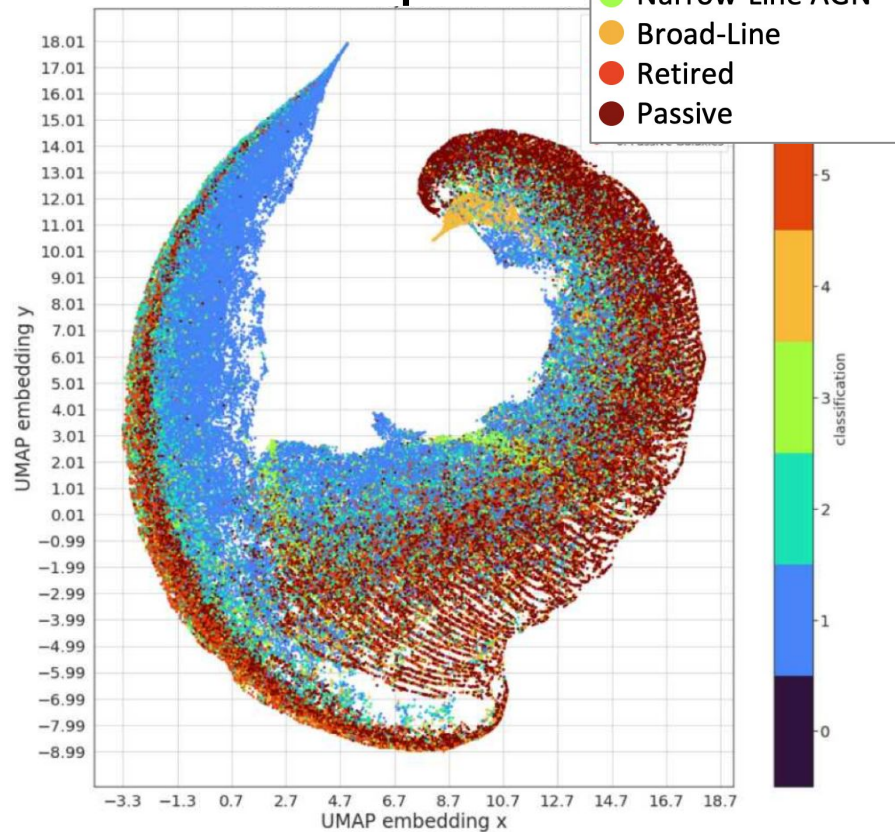
Classes

- Star Forming
- Composite
- Narrow-Line AGN
- Broad-Line
- Retired
- Passive

(Left) Star-Forming, Composite and Narrow-line AGN; **(Middle)** BROADLINE flag is set by SDSS pipeline \rightarrow Broad-line AGN; **(Right)** Retired & Passive galaxies

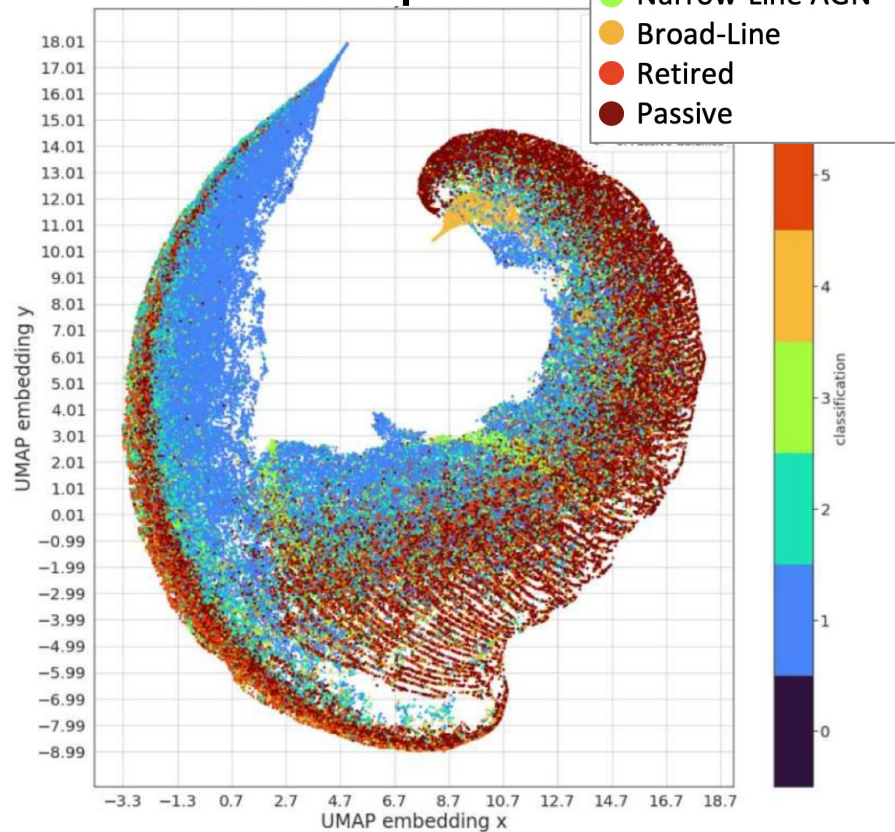
WHAN: Equivalent width of H α vs. [NII]/H α ratio (Cid-Fernandes et al. 2011)

UMAP from spectra

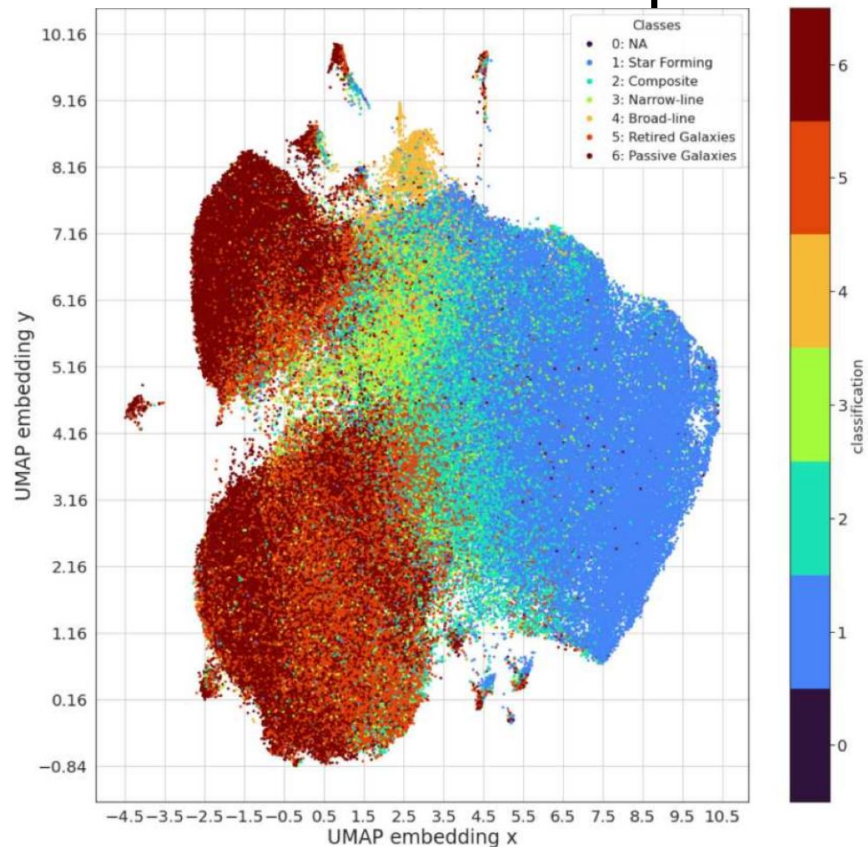


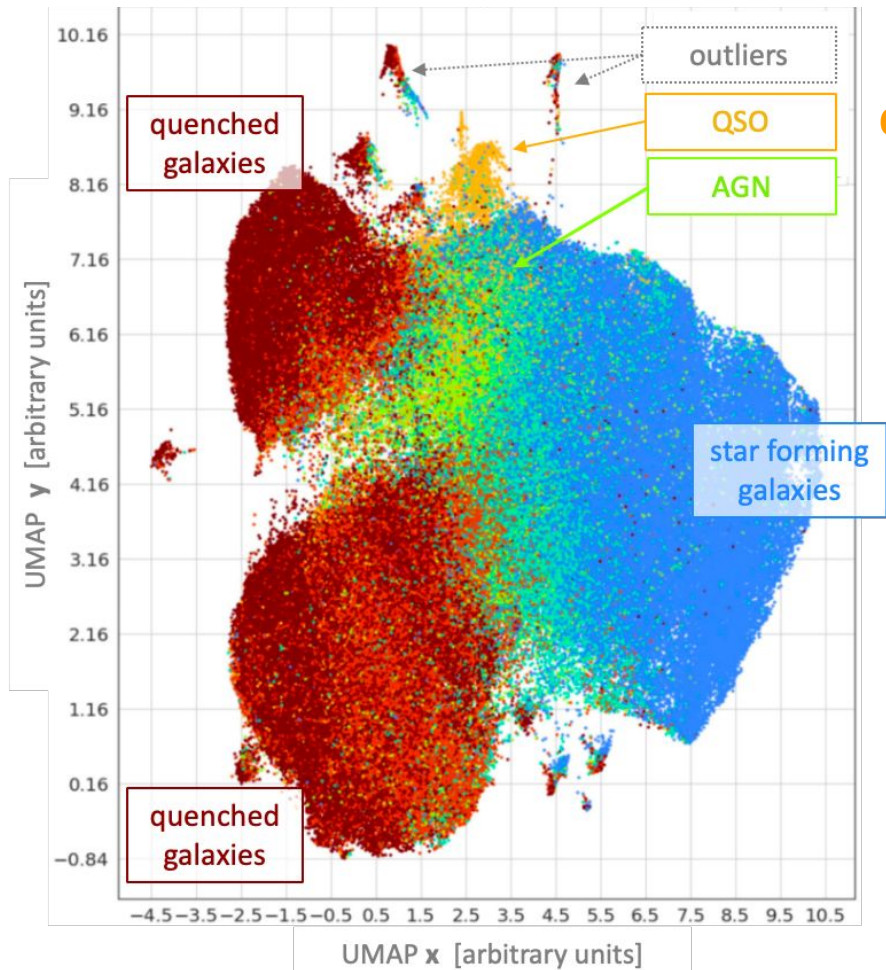
(Pat, Juneau, Böhm et al., 2022)

UMAP from spectra

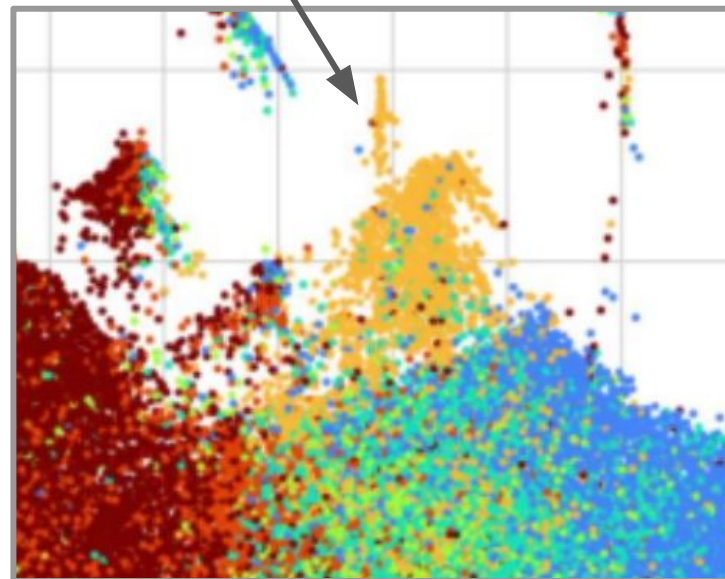


UMAP from AE latent space



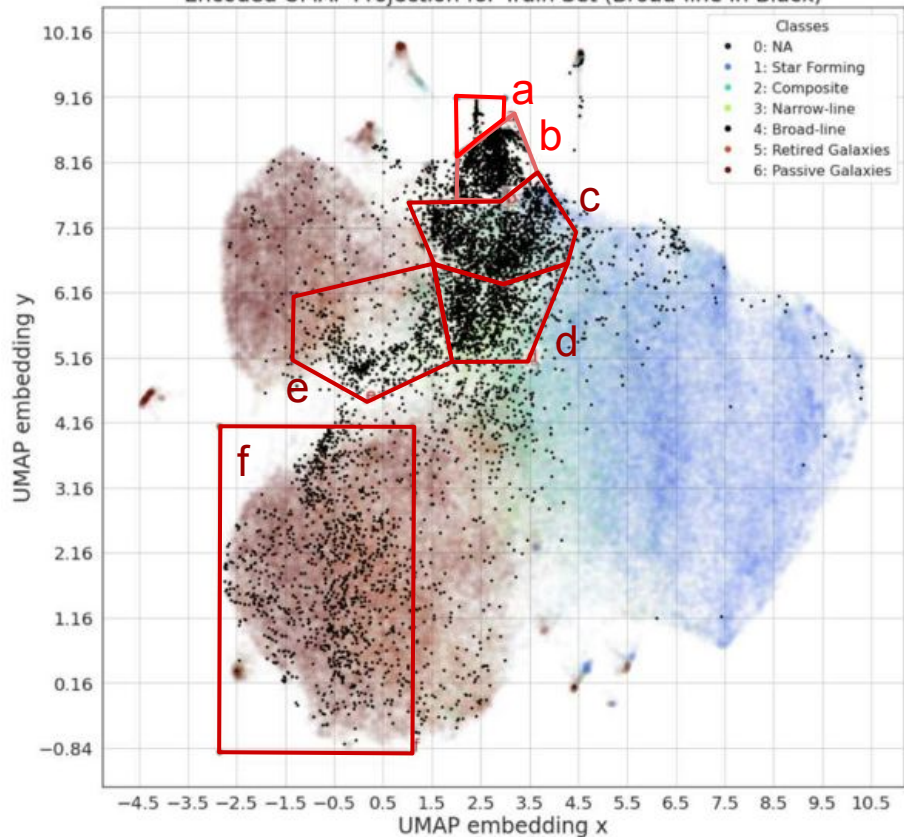


QSO (BROADLINE label)

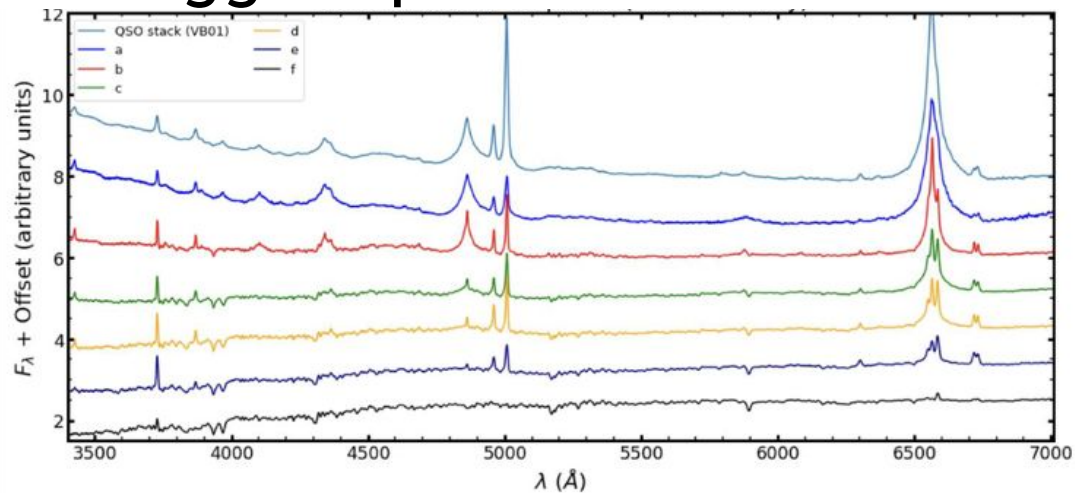
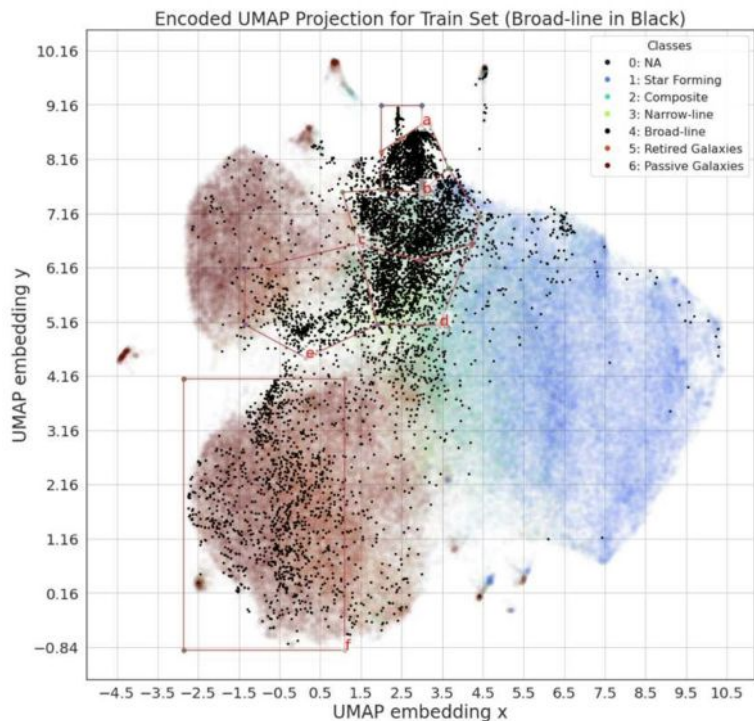


Trends among BROADLINE flagged spectra

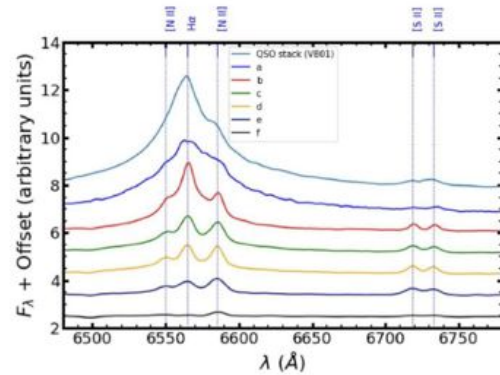
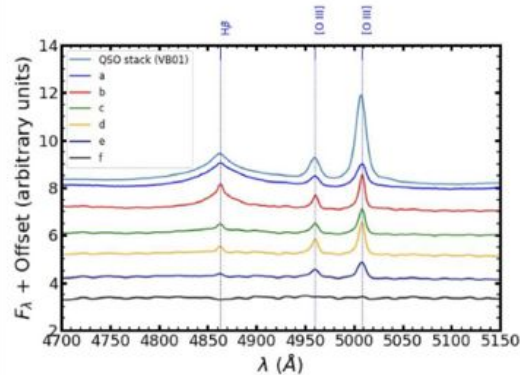
Encoded UMAP Projection for Train Set (Broad-line in Black)



Trends among BROADLINE flagged spectra

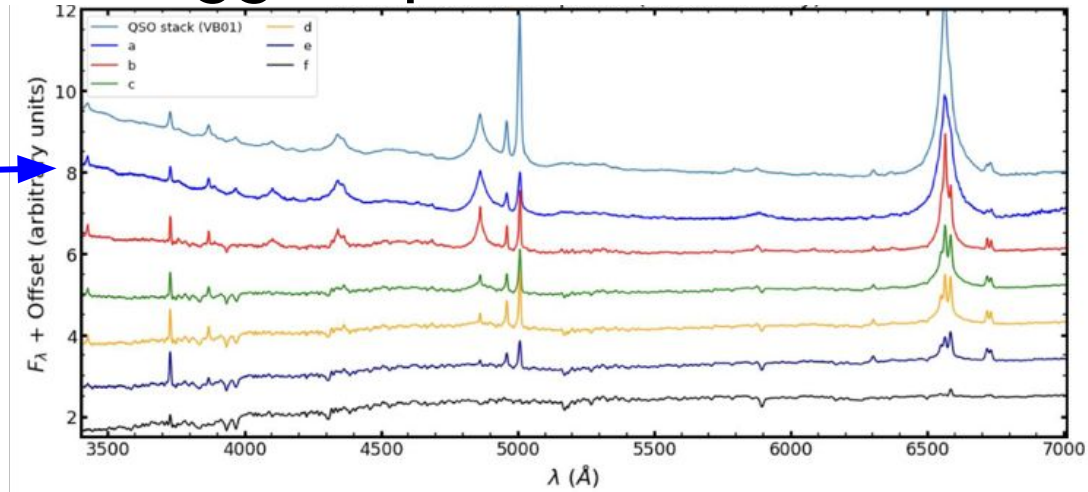


(a) Stacked spectra of broad-line sequence from UMAP (Figure 6)

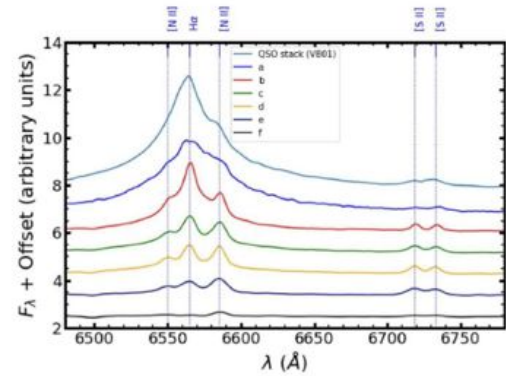
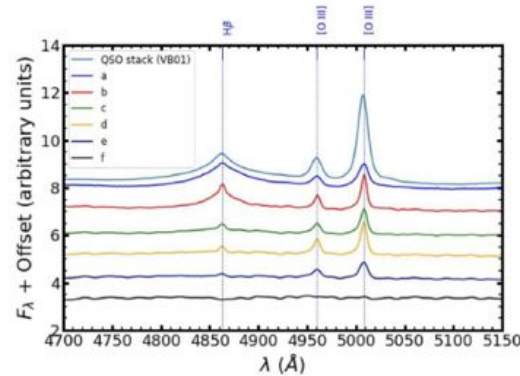


Trends among BROADLINE flagged spectra

The distinct branch consists of QSO spectra nearly identical to average SDSS quasar (VandenBerk 2001)



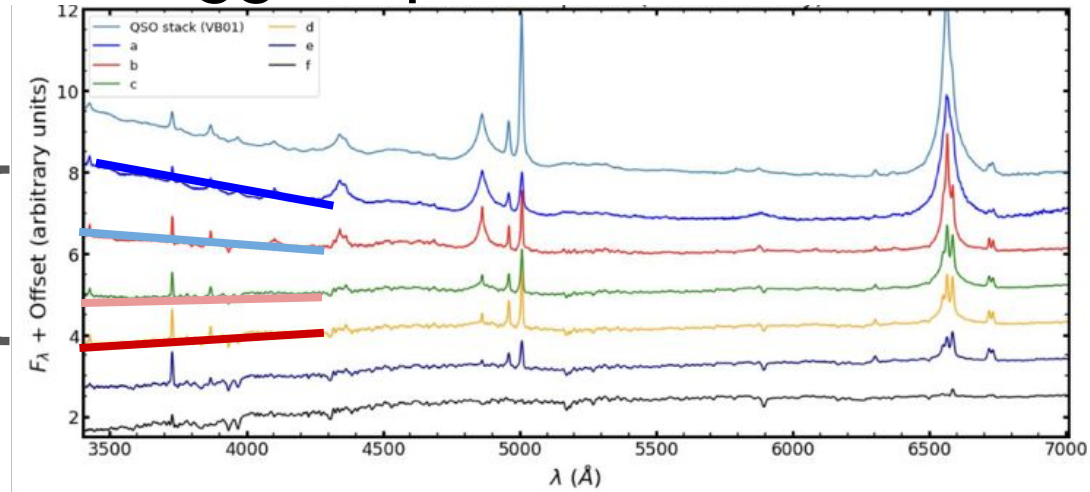
(a) Stacked spectra of broad-line sequence from UMAP (Figure 6)



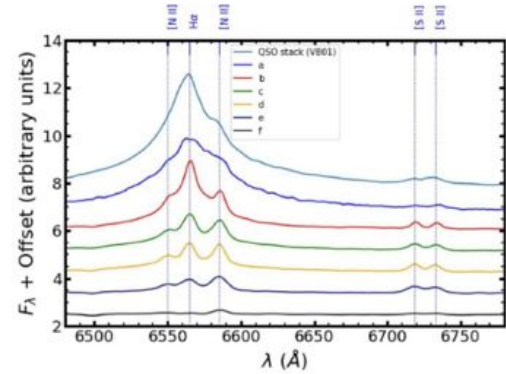
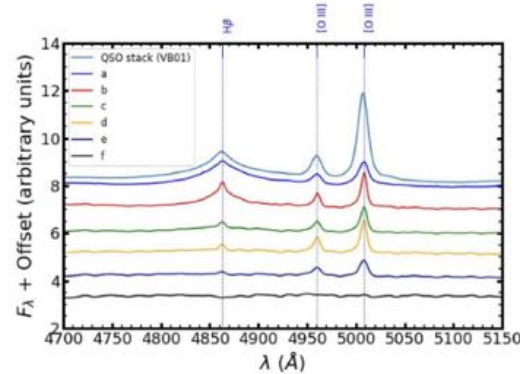
Trends among BROADLINE flagged spectra

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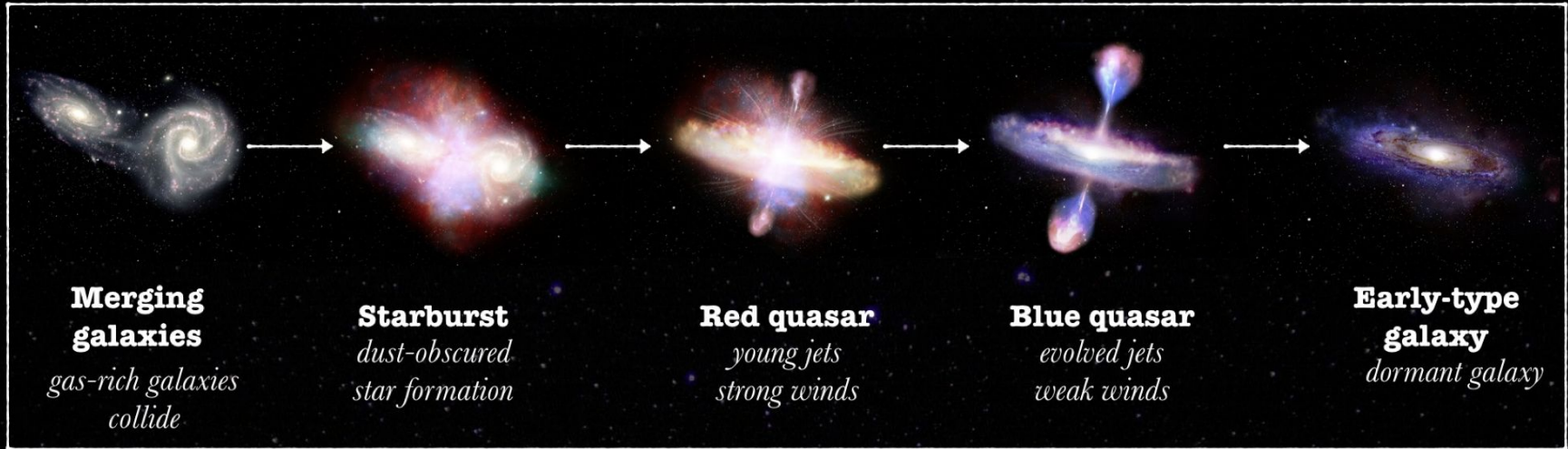
Then a trend of reddening/obscuration



(a) Stacked spectra of broad-line sequence from UMAP (Figure 6)



Red quasars as a rare population (or in a special phase)



- DESI secondary program led by Vicky Fawcett ([Fawcett et al. 2023](#))
- Construct a more complete and diverse census of quasars, particularly reddened and extremely reddened quasars.
- Test whether dust-obscured quasars represent a special (transition) phase in the evolution of quasars and the formation of massive galaxies.



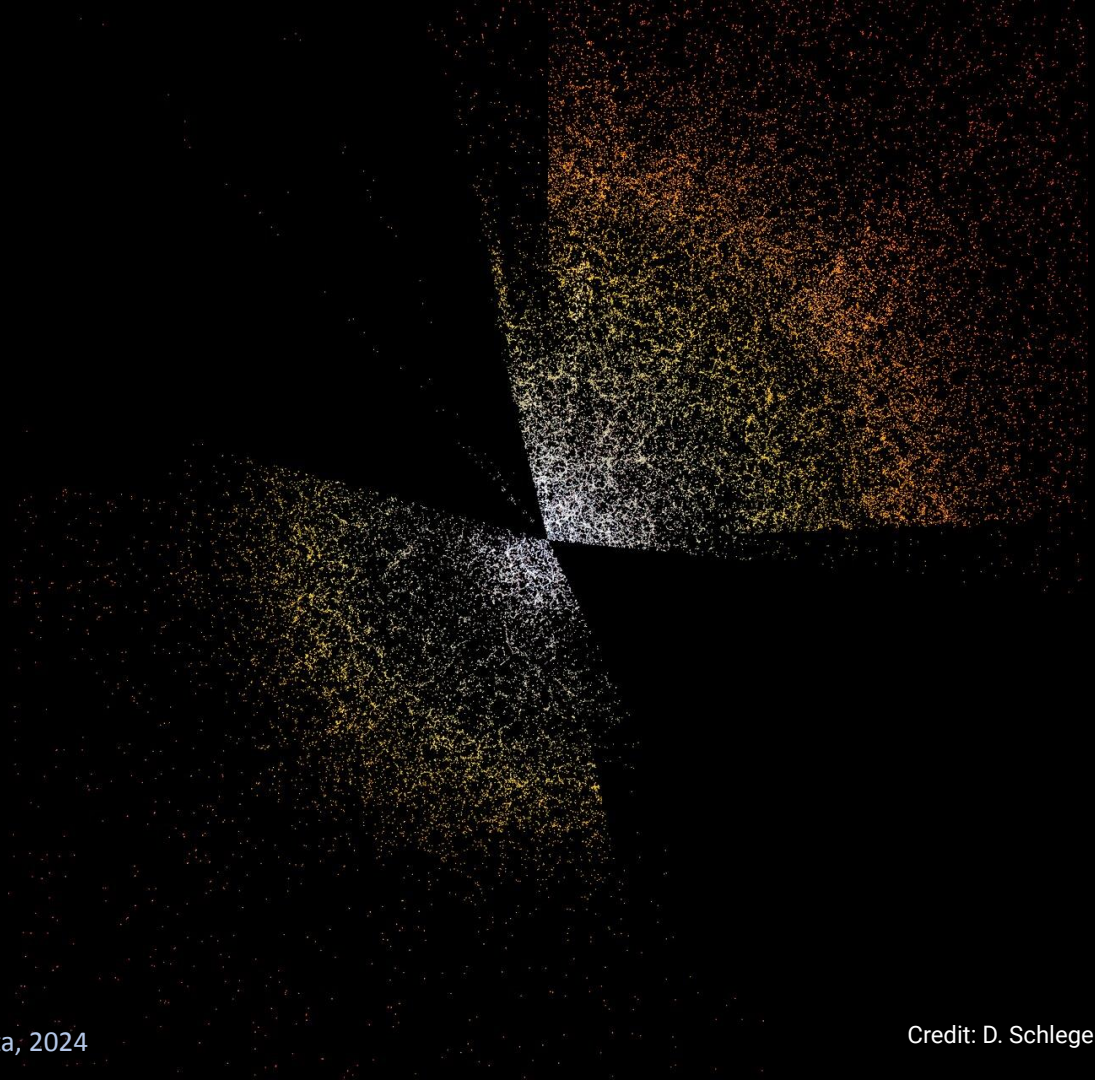
DESI (Dark Energy Spectroscopic Instrument)

Installed at the 4-meter Mayall telescope (Kitt Peak, AZ)

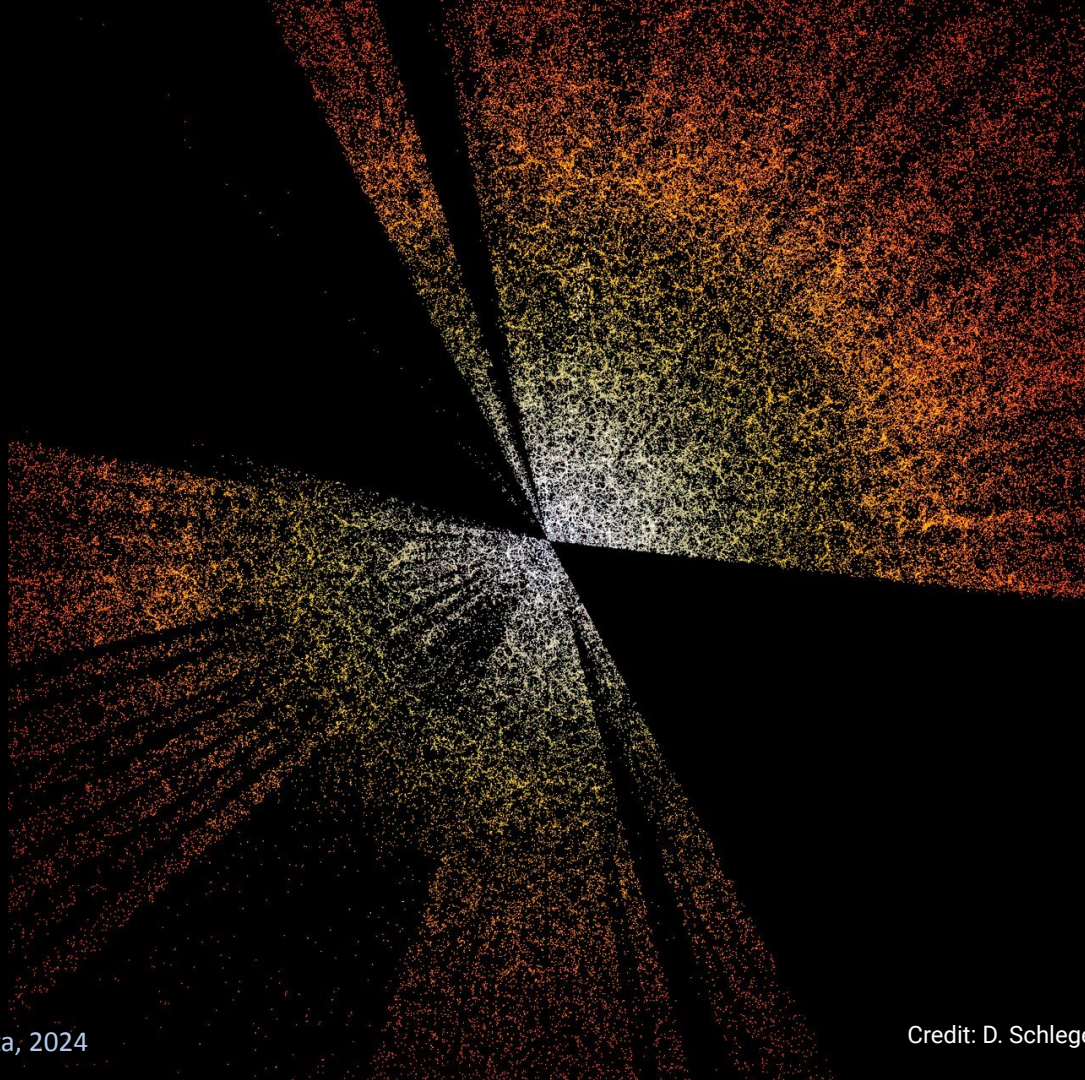
Will obtain spectra of **40 million galaxies & quasars** in 5 years (2021-2026)



From SDSS...



...to DESI



Embarking on a Quasar rescue mission

DESI Bright Galaxy Survey:

- Bright ($r < 19.5$) + Faint ($r < 20.3$ AB)
- Star/Galaxy separation ($G - r > 0.6$) removes stars (good) and some quasars (sad)

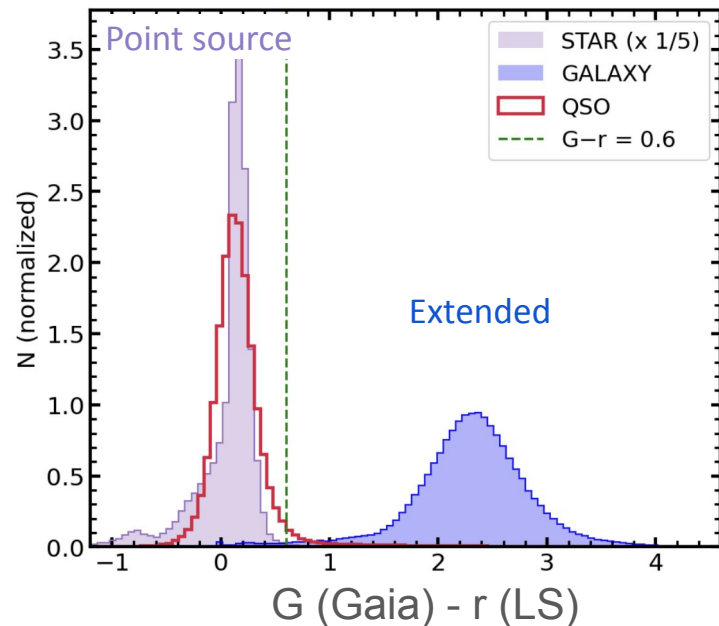
Mission:

- Put the quasars back in!
(without increasing stellar contamination)

Strategy:

- Color cuts leveraging WISE bands probing AGN-heated dust ($g, r, z, W1, W2$)

Star/Galaxy separation



Embarking on a Quasar rescue

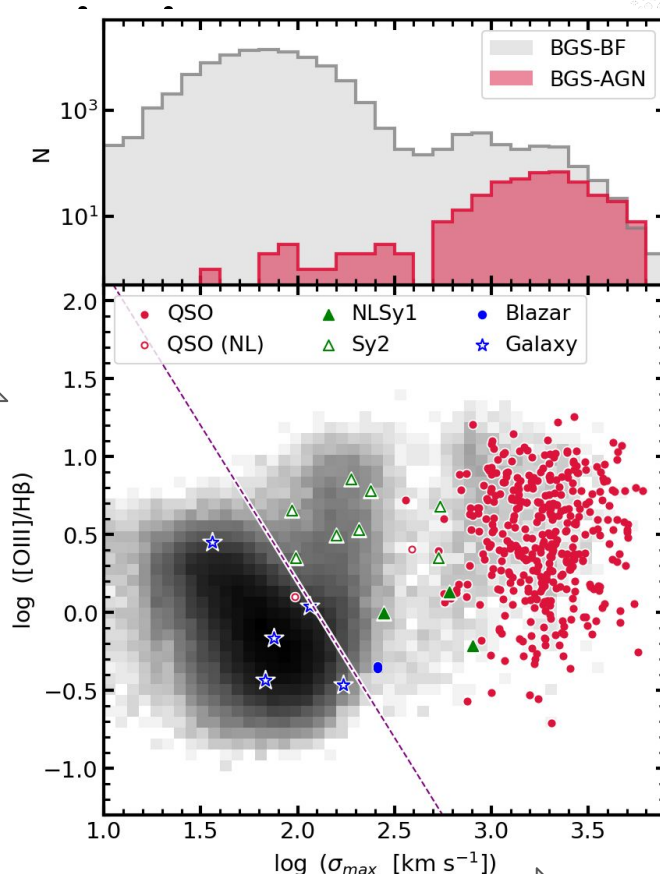
Final results

- As anticipated, BGS AGN targets are mostly Quasars but often with some host galaxy light → intermediate between the BGS sample and the QSO sample
- Also found: Type 2 Seyfert (Sy2) AGN, Narrow-Line Seyfert 1 (NLSy1) AGN, Blazars, BAL quasars, red quasars

DESI DR1: >15,000 unique BGS AGN targets

Will be part of AGN/QSO Catalog (millions)

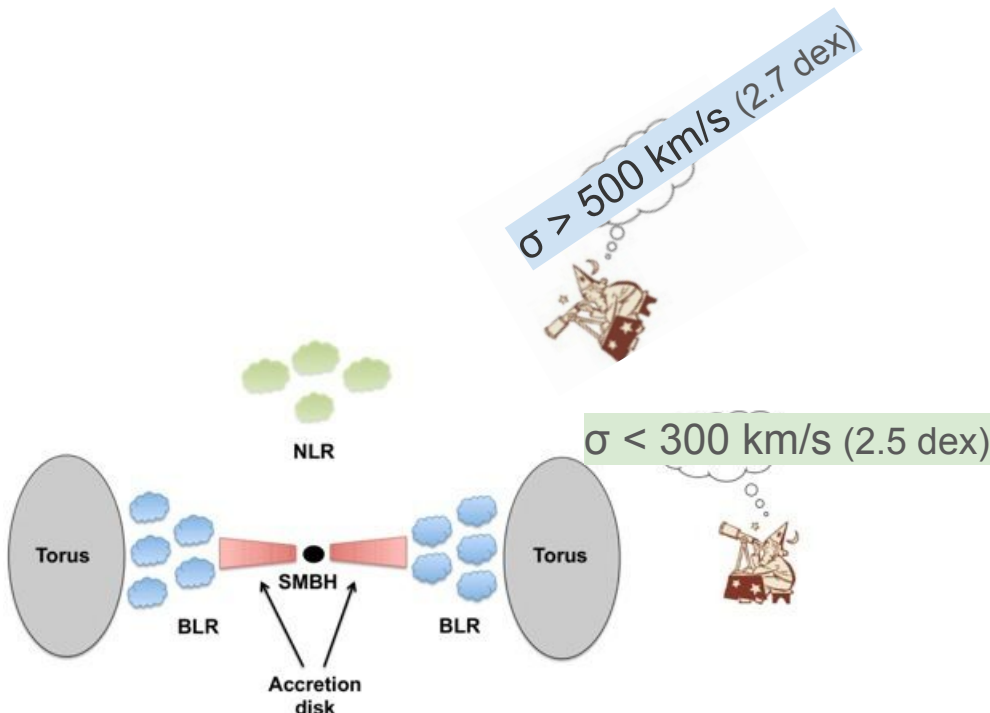
↑
gas excitation



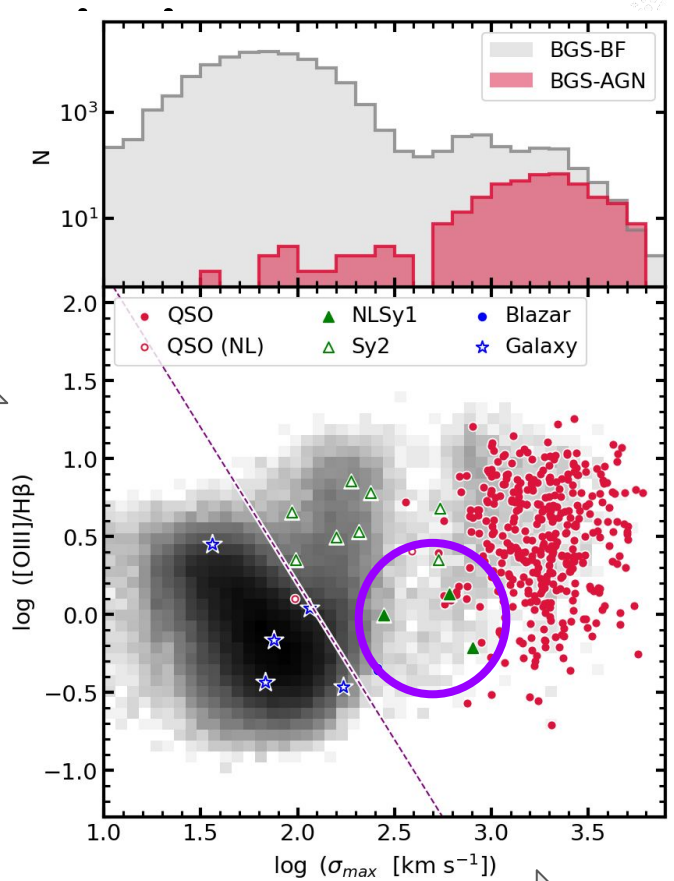
→
gas motion

(Juneau et al. 2024; arXiv:2404.03621)

Embarking on a Quasar rescue

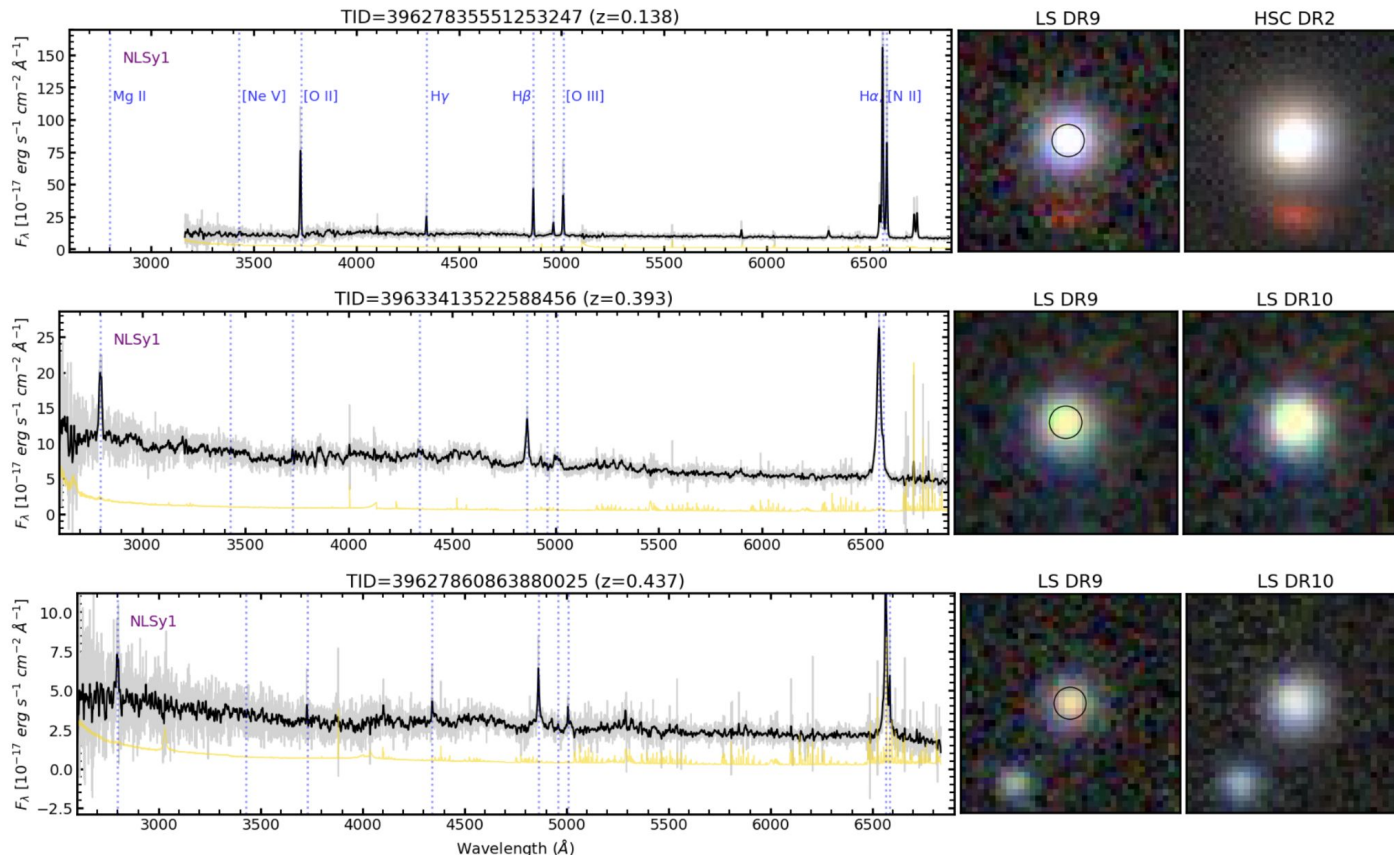


gas excitation



gas motion

Spectra + images for NLSy1 = Narrow-Line Seyfert 1



Seyfert 1 \rightarrow normally have *broad* lines

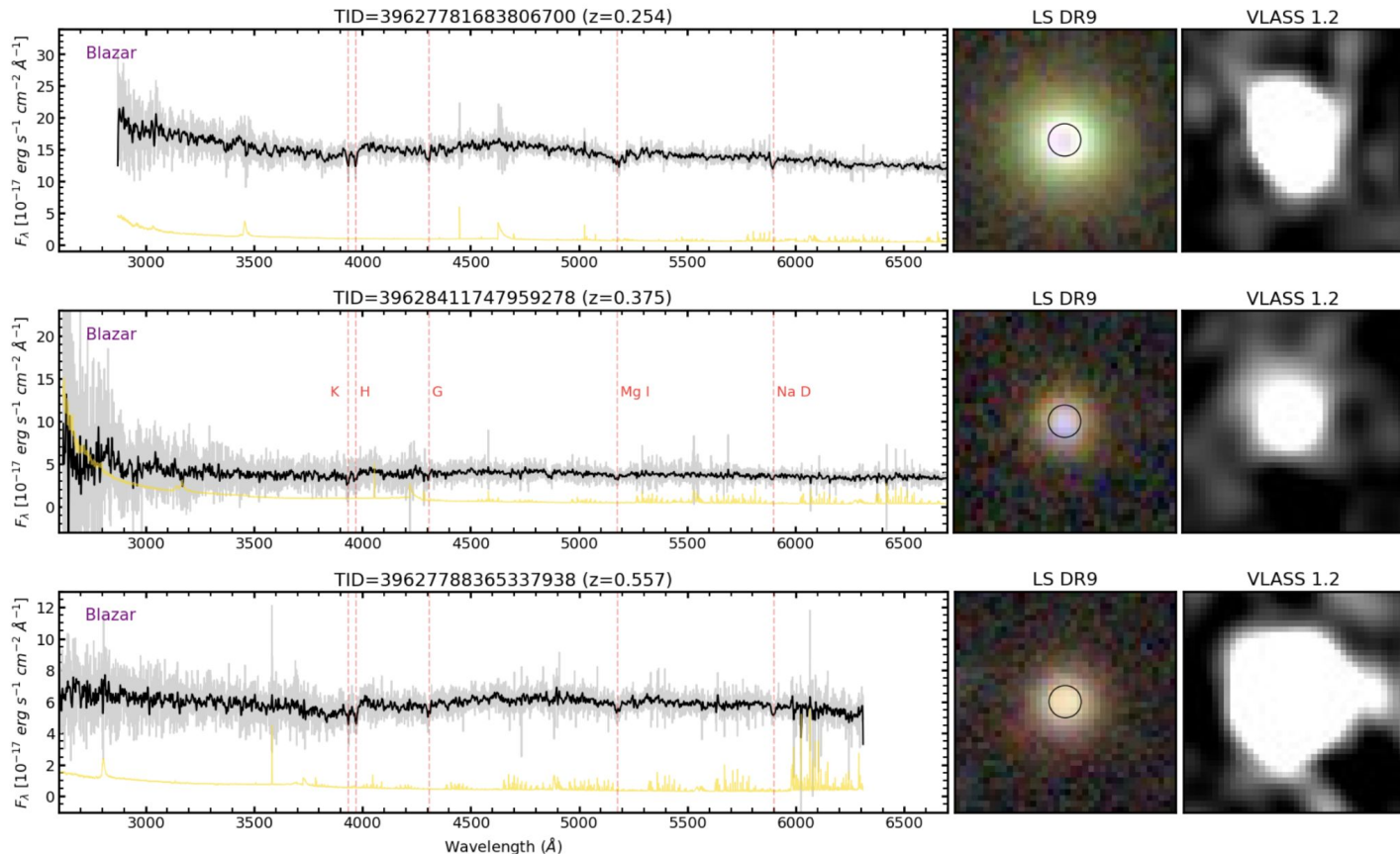
Scenario:

Young AGN with a low-mass black hole growing fast

\rightarrow analogs to high-redshift AGN found with JWST?

(topic: BH growth / BH seeds)

Spectra + images for Blazars



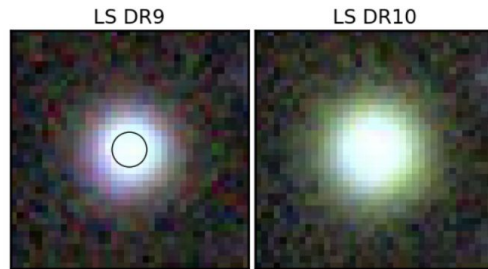
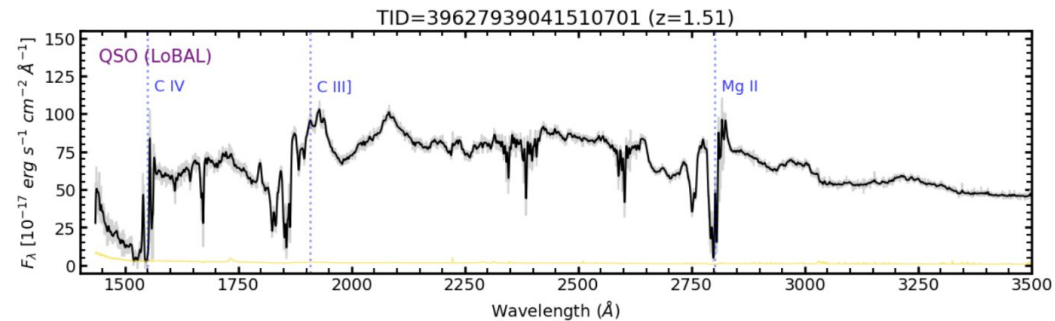
spectype=GALAXY

Stellar absorption lines but continuum rising toward the blue!

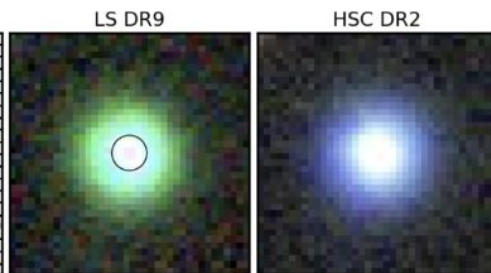
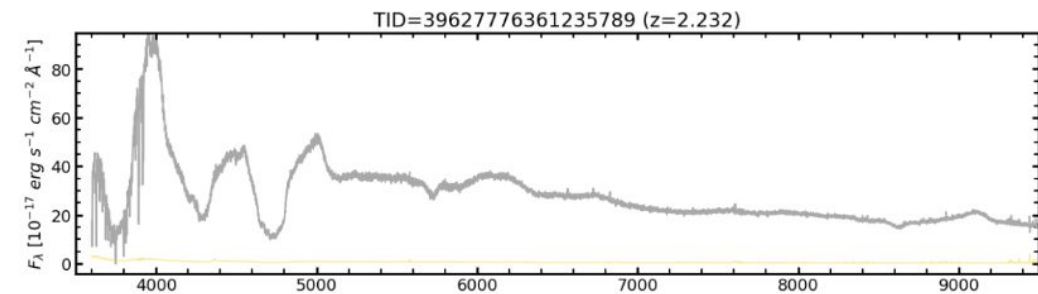
Obvious VLASS (3 GHz) counterparts

→ Looking toward radio jet (rare orientation)

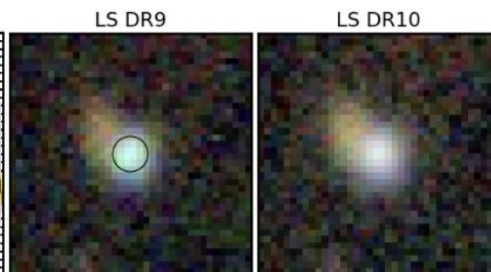
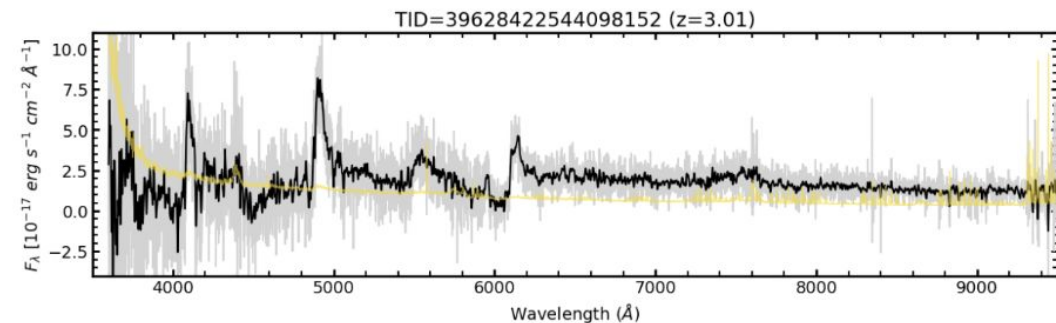
Spectra + images for BAL quasars



Missed by RR
(identified by VI)



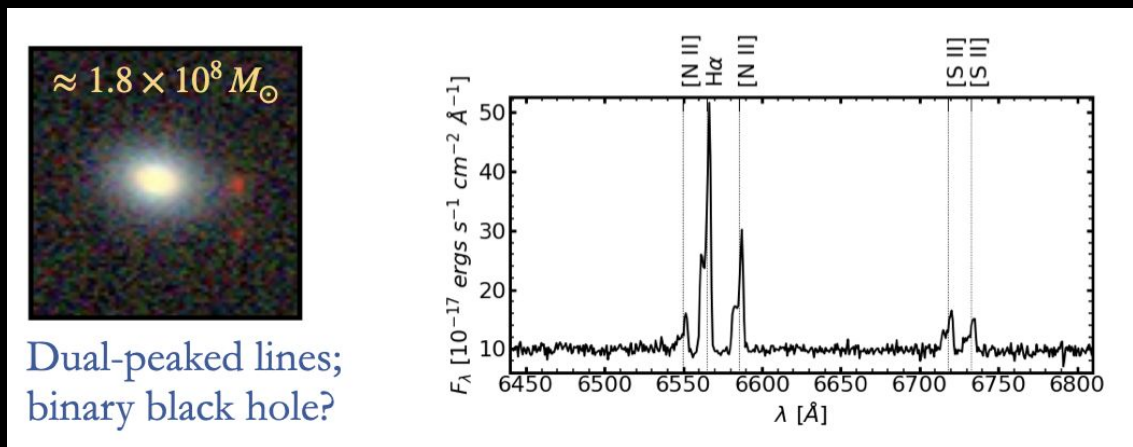
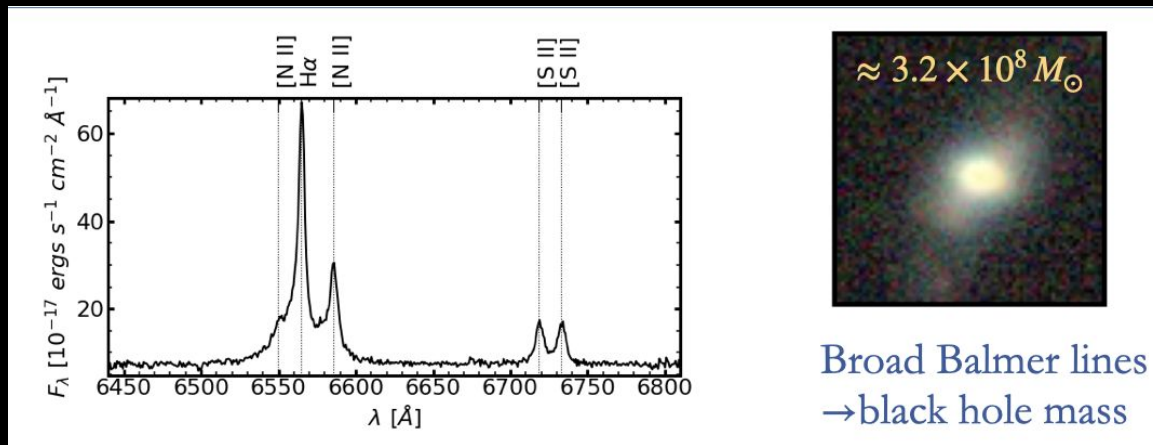
Bright ($r_{AB} < 17.5$)
PSF morphology
*also QSO target



Faint ($r_{AB} > 17.5$)
non-PSF morph.
due to projected
neighbor?

Active black holes in low-mass dwarf galaxies

Ragadeepika Pucha (University of Utah)



Research done using
NOIRLab/Data Lab
and NERSC

Conclusions

- Probabilistic Autoencoders (PAE) can:
 - accurately reconstruct spectra (noiseless and inpainted over gaps)
 - find anomalies (with or without pre-assigned labels)
- Combining PAE + UMAP reveals interesting sequences worth studying further (Pat, Juneau et al. 2022)
 - Different populations (or evolutionary phases) of quasars
 - Similar to using VAE (Portillo et al. 2020) & SPENDER (Melchior et al. 2023 [talk], Liang et al. 2023a,b [poster])
- Developed target selection to recover *missing* quasars in the DESI Bright Galaxy Survey → found normal quasars but also some rare gems (Juneau et al. 2024)
- **Discussion:** need to strategize about how to select targets but can also exploit some contaminants/failures as they can reveal interesting objects!



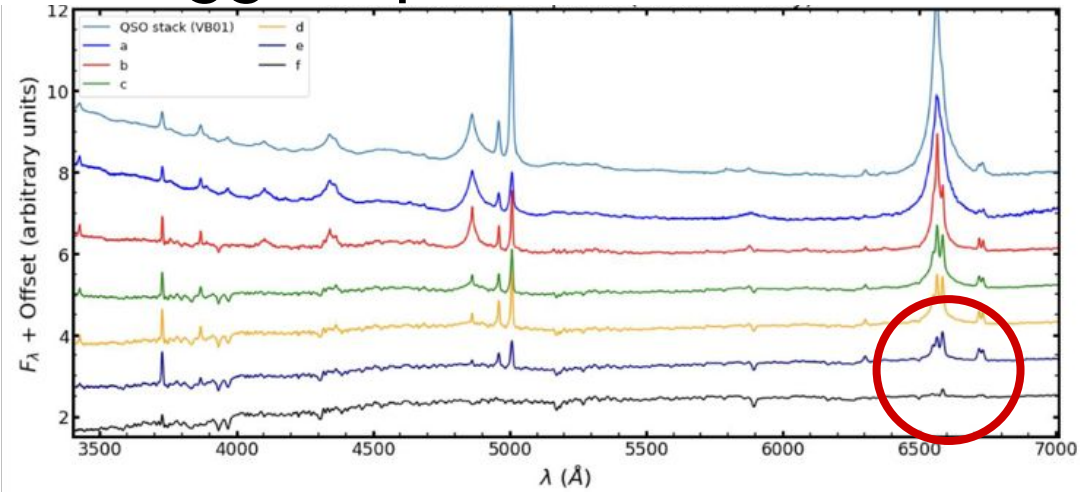
S. Juneau – Rare Gems in Big Data, 2024

Trends among BROADLINE flagged spectra

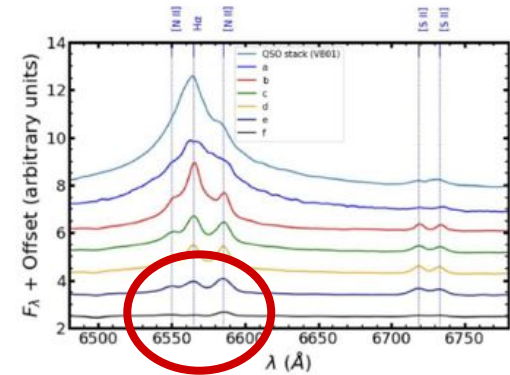
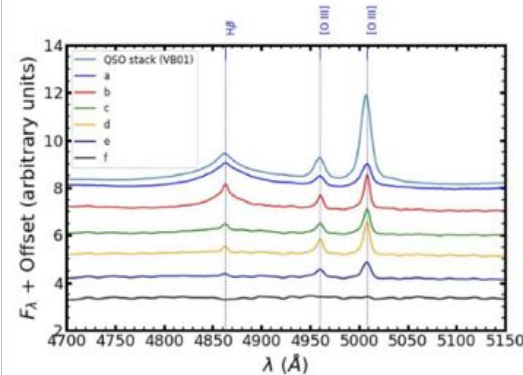
The distinct branch consists of QSO spectra virtually identical to average SDSS quasar (VandenBerk 2001)

Then a trend of reddening/obscuration

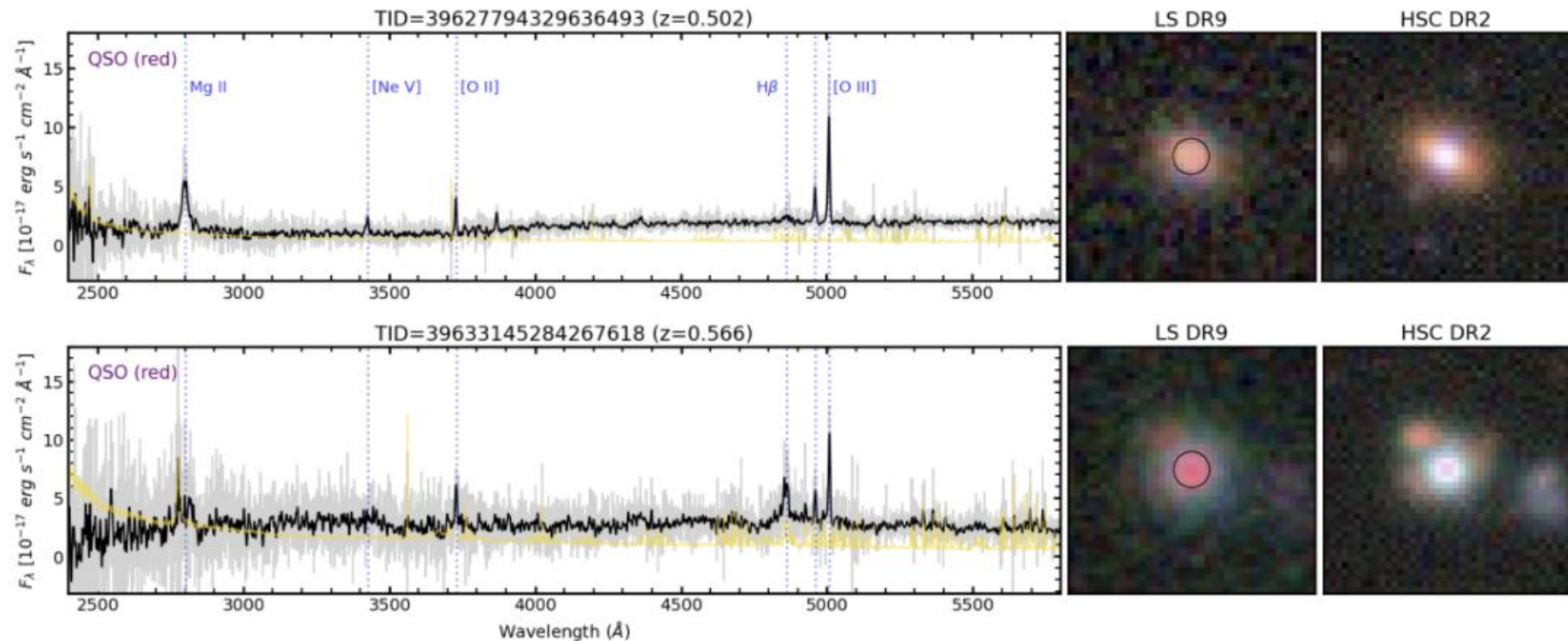
Outliers don't have *true* broad (Balmer) lines and arise from SDSS definition (FWHM > 400 km/s) picking up LINERs in massive galaxies



(a) Stacked spectra of broad-line sequence from UMAP (Figure 6)



Spectra + images for red quasars



Probabilistic Autoencoder (PAE)

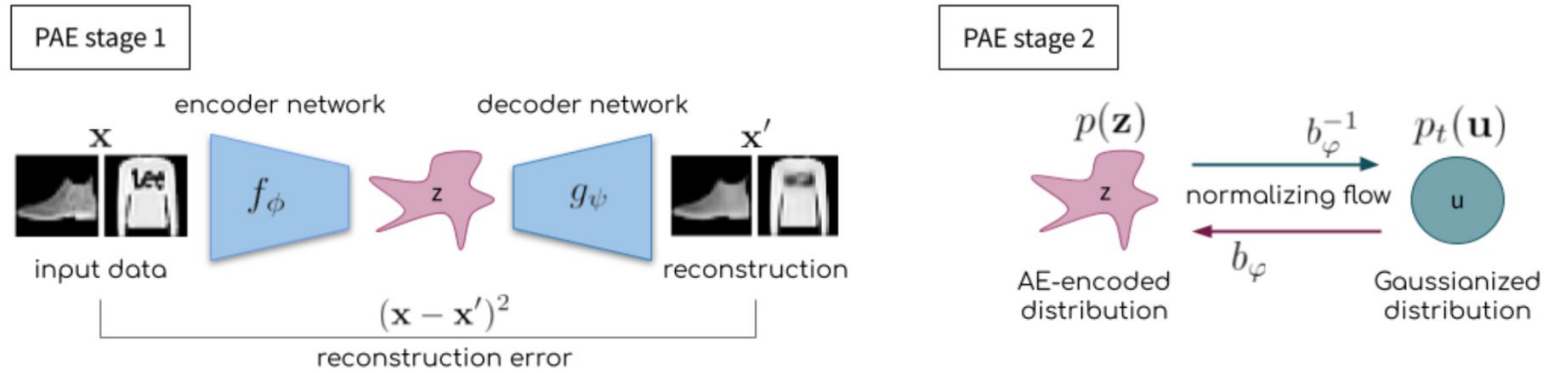


Figure 1. The two-step training of a Probabilistic Autoencoder (PAE) illustrated on an image data set. In the first stage, an autoencoder (AE) consisting of two neural networks is trained to minimize the reconstruction error after compressing (encoding) the data in a lower-dimensional latent space and decompressing (decoding) it back into the high-dimensional data space. In the second stage, a normalizing flow (NF) is trained to learn a bijective mapping from the AE-encoded latent space to a space in which the encoded data follows a Gaussian distribution.

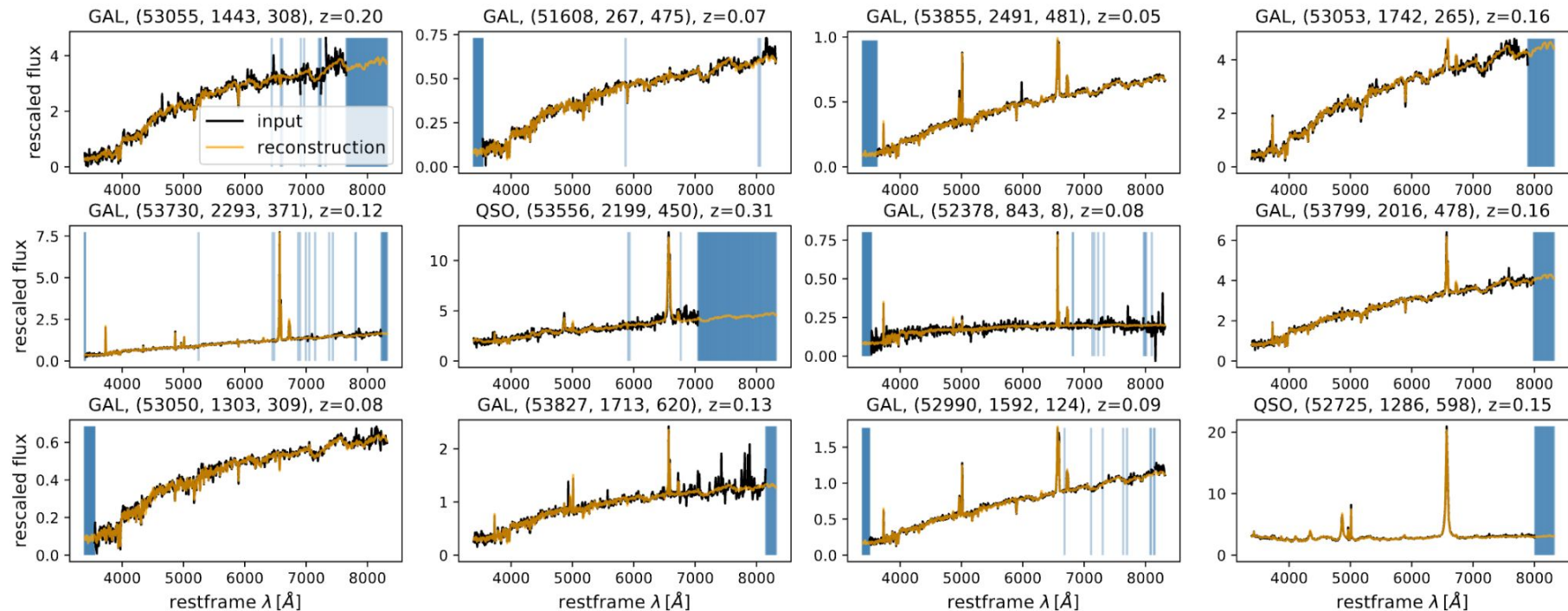
<https://github.com/VMBoehm/PytorchPAE>

(Böhm & Seljak 2022; Böhm, Kim & Juneau 2023)

Example PAE-reconstructed SDSS spectra

Black: observed spectra (noisy with masked regions)
Orange: denoised and inpainted reconstructed spectra

CLASS (MJD, PLATE, FIBERID)



Probabilistic Autoencoder (PAE)

- Strategy to account for certain classes (e.g., QSO) being more rare than others (e.g., GALAXY)
 - Evaluation conditional probability given a label (class+subclass)
- Compare without (e.g., Pat+22, Liang+23a) and with (Böhm+23) conditional density estimation

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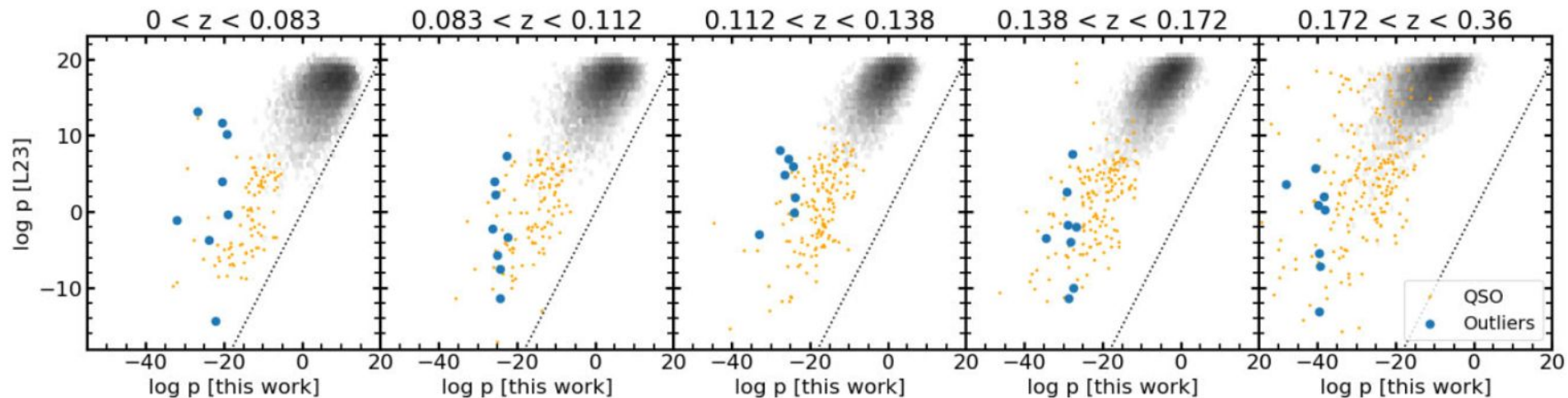
(Böhm & Seljak 2022; Böhm, Kim & Juneau 2023)

Comparing PAE with/without conditional density

Black: quiescent galaxies

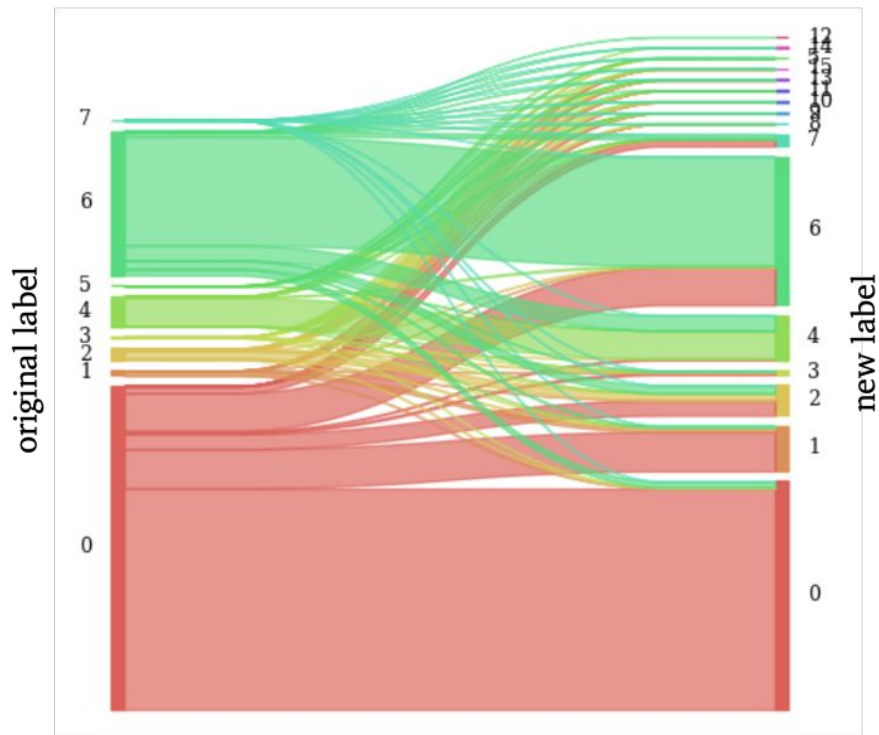
Blue: outliers in each redshift bin

Orange: QSO spectra (conditional vs. not)

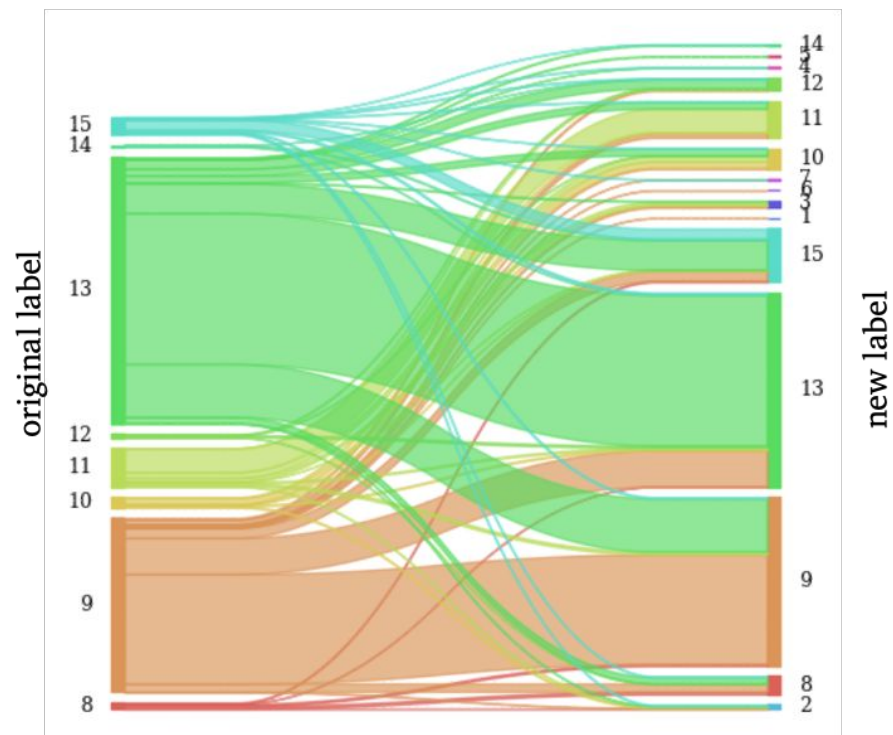


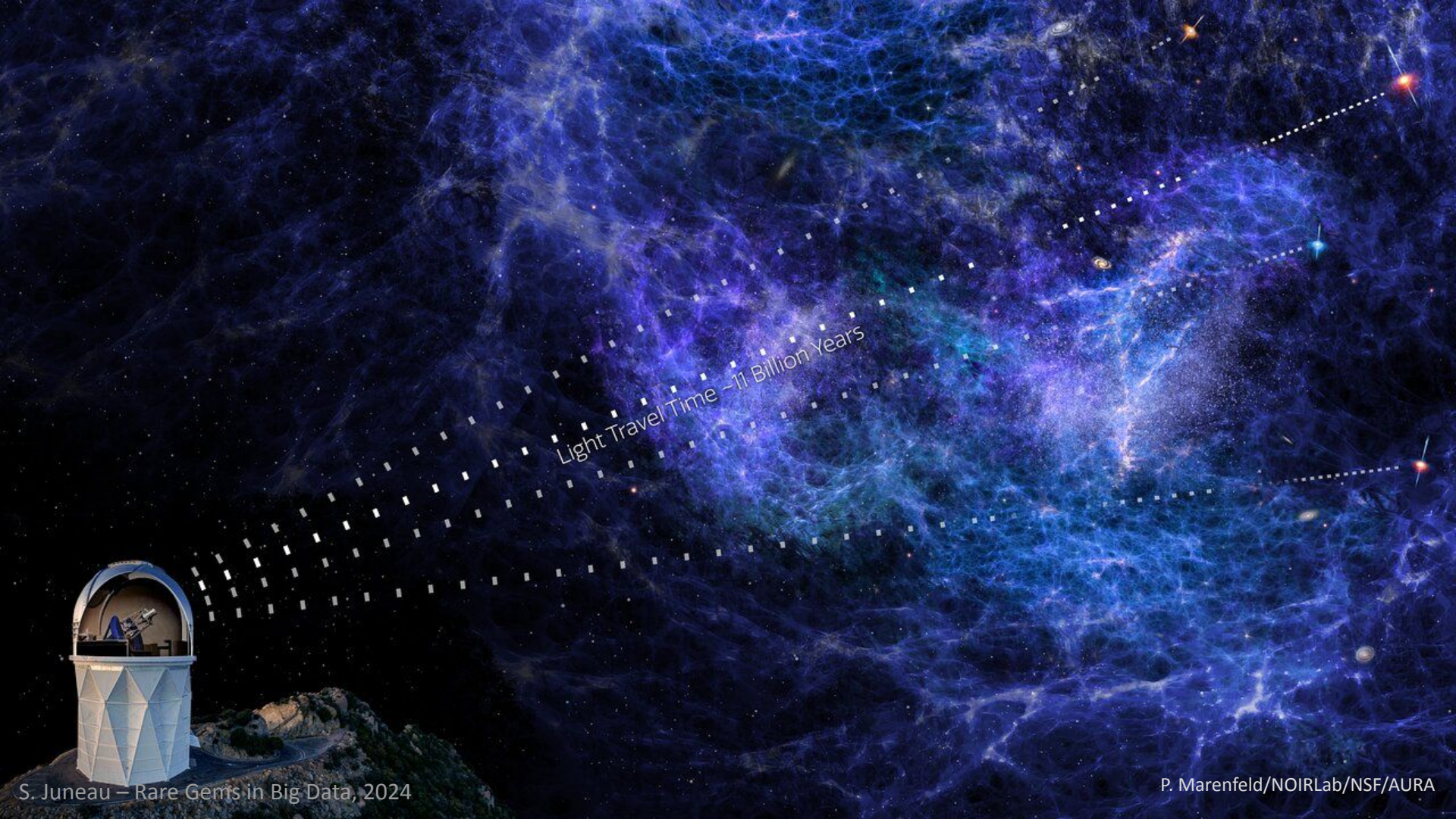
Relabeling after one round of AE

CLASS = GALAXY



CLASS = QSO





Light Travel Time - 11 Billion Years



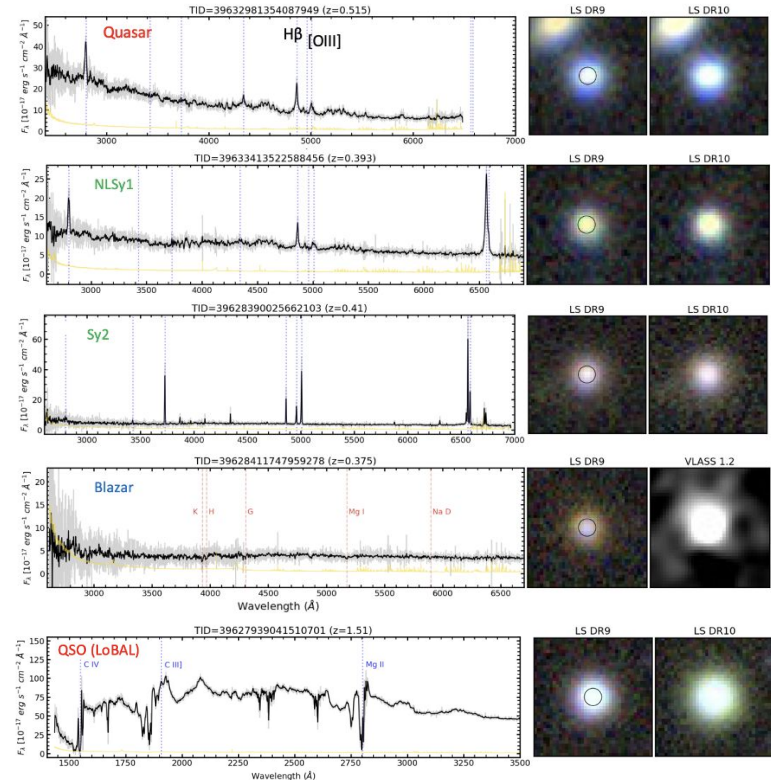
Embarking on a Quasar rescue mission

Final results

- As expected, BGS AGN targets are mostly Quasars but often with some host galaxy light \rightarrow intermediate between the BGS sample and the QSO sample
- Also found: Type 2 Seyfert (Sy2) AGN, Narrow-Line Seyfert 1 (NLSy1) AGN, Blazars, BAL quasars, red quasars

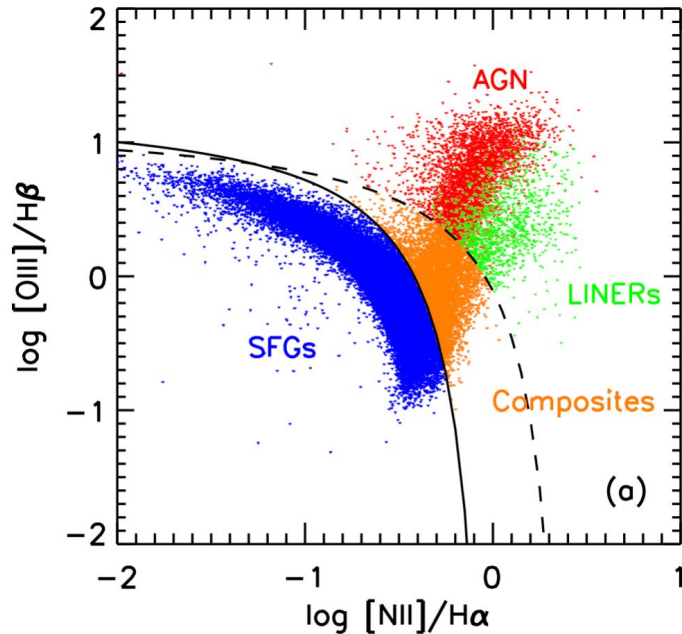
DESI DR1: >15,000 unique BGS AGN targets

Will be part of AGN/QSO Catalog (millions)



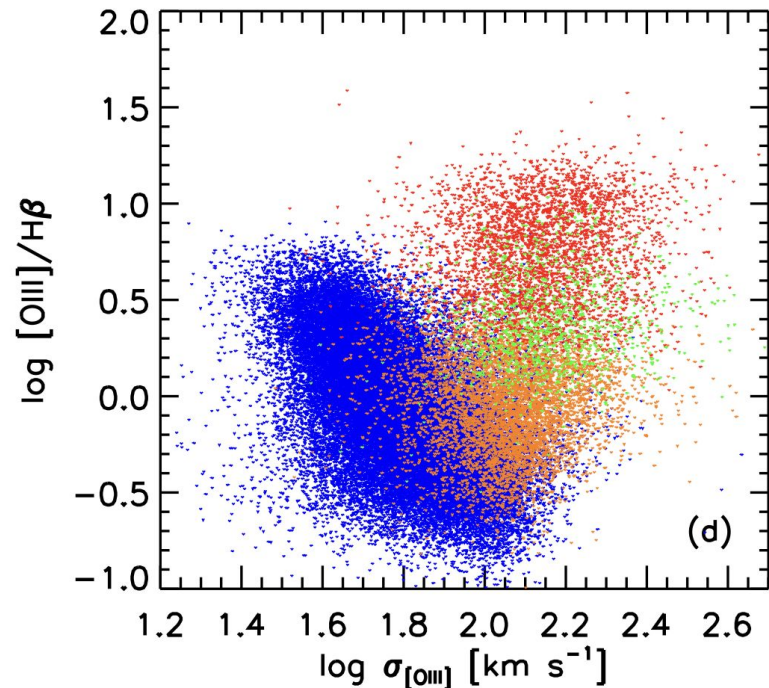
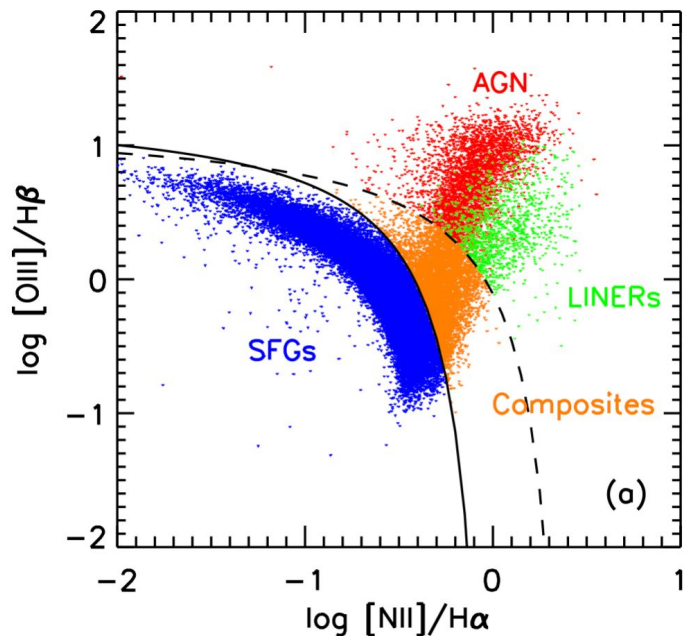
Kinematic-Excitation (KEx) diagram

- Calibrated at low redshift with SDSS using BPT (below) and VO87 diagrams



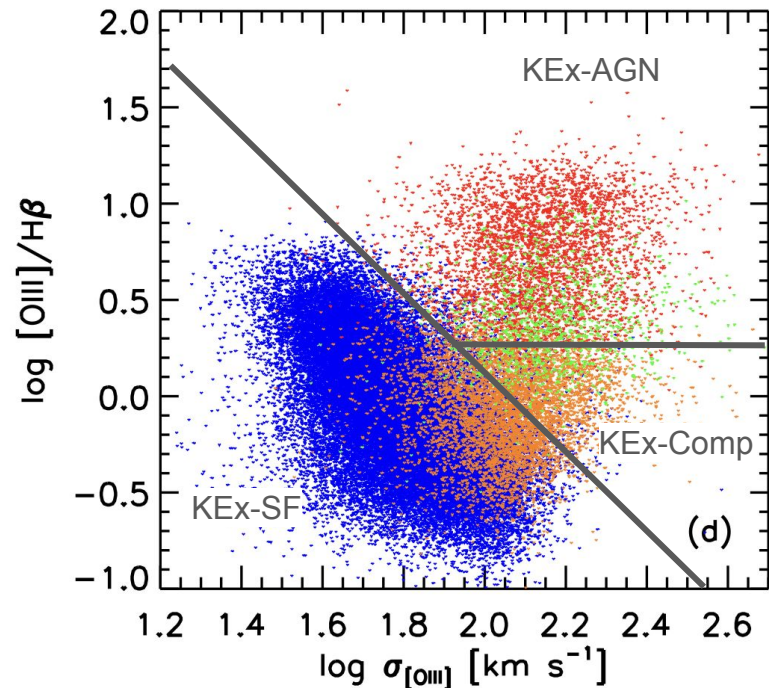
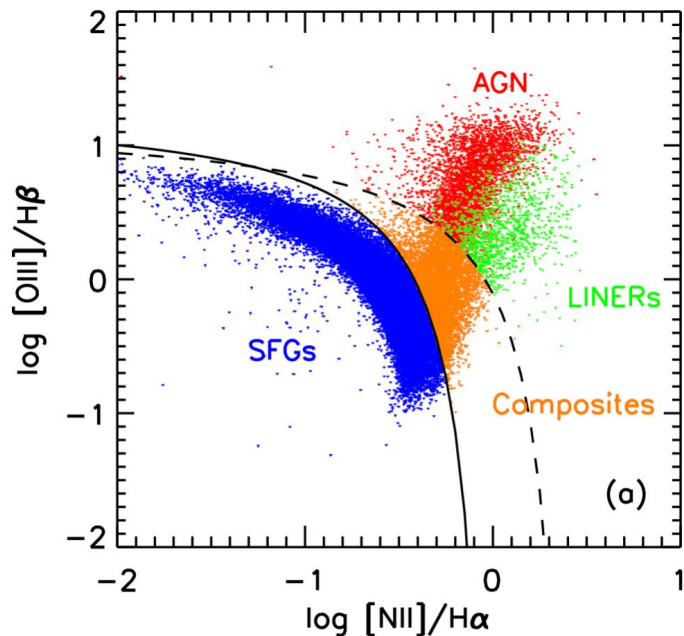
Kinematic-Excitation (KEx) diagram

- Calibrated at low redshift with SDSS using BPT (left) and VO87 diagrams
- Uses the linewidth of [OIII]5007 as horizontal axis
- Can be applied to $z \sim 1$ with optical spectra



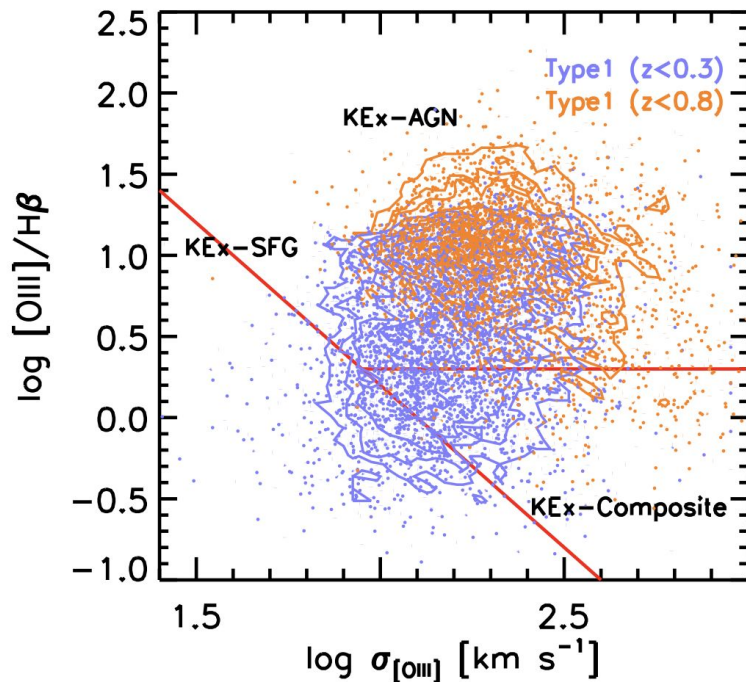
Kinematic-Excitation (KEx) diagram

- Calibrated at low redshift with SDSS using BPT (left) and VO87 diagrams
- Uses the linewidth of [OIII]5007 as horizontal axis
- Can be applied to $z \sim 1$ with optical spectra



Kinematic-Excitation (KEx) diagram

- But what about Type 1 (Broad-Line) AGN?



- Low-z sample (blue) reaches to lower AGN luminosities → extends into SF region
- High-z sample (red) biased toward higher AGN lum → cleaner separation

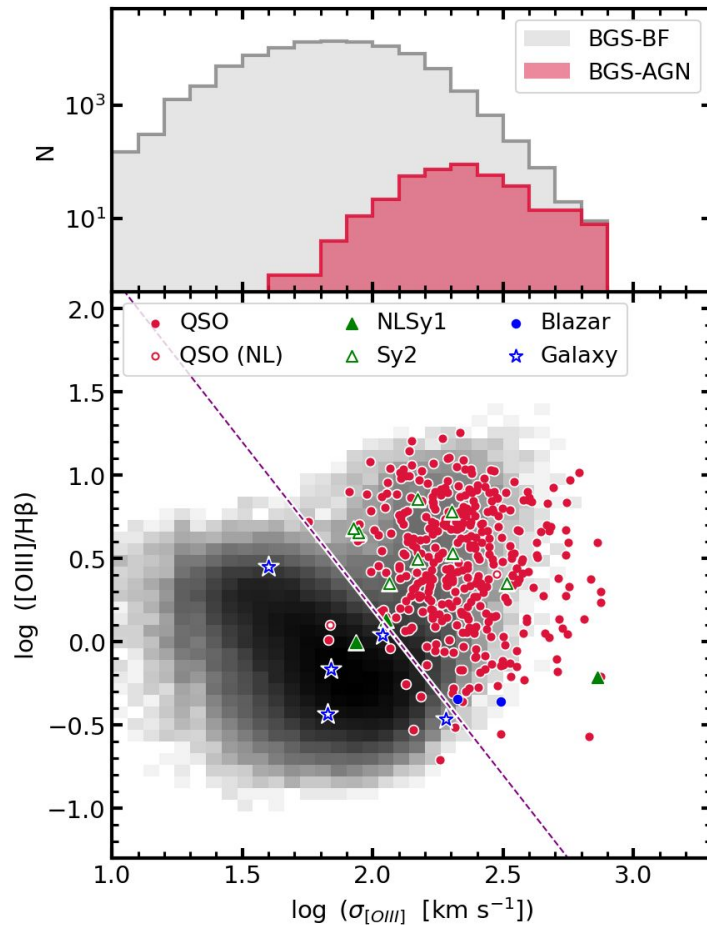
But...

... we can split them further thanks to their broad linewidth in $H\beta$!

KEx diagram

- $[\text{OIII}]/\text{H}\beta$ is the ratio of the *narrow* line fluxes (caveat: fastspecfit assumes single Gaussian components for $[\text{OIII}]5007$ so no outflow)
- $\sigma([\text{OIII}])$ in km/s

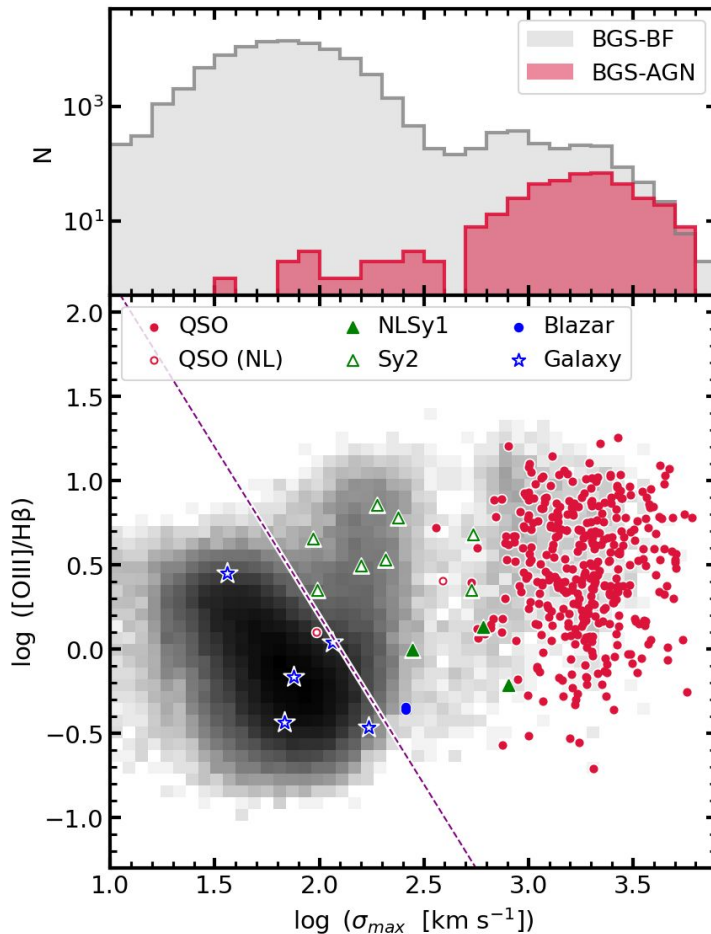
→ no distinction b/w Type 1 vs. Type 2 AGN



Modified KEx diagram

- $[\text{OIII}]/\text{H}\beta$ is the ratio of the *narrow* line fluxes (caveat: fastspecfit assumes single Gaussian components for $[\text{OIII}]5007$ so no outflow)
- Replace $\sigma([\text{OIII}])$ with $\max\{\sigma(\text{MgII}, \text{H}\beta, \text{H}\alpha)\}$

→ better visualize the Type 1 vs. Type 2 AGN dichotomy & relative proportions



Modified KEx diagram

Side-by-side comparison

Left: $\sigma([\text{OIII}])$

Right: $\max\{\sigma(\text{MgII}, \text{H}\beta, \text{H}\alpha)\}$

$[\text{OIII}]/\text{H}\beta$ ratio is for the narrow lines only (i.e., excluding broad $\text{H}\beta$)

