doi:10.1093/mnras/staa3259



Constraining photoionization models with a reprojected optical diagnostic diagram

Xihan Ji [®]★ and Renbin Yan★

Department of Physics and Astronomy, University of Kentucky, 505 Rose Street, Lexington, KY 40506, USA

Accepted 2020 October 15. Received 2020 October 14; in original form 2020 July 20

ABSTRACT

Optical diagnostic diagrams are powerful tools to separate different ionizing sources in galaxies. However, the model-constraining power of the most widely used diagrams is very limited and challenging to visualize. In addition, there have always been classification inconsistencies between diagrams based on different line ratios, and ambiguities between regions purely ionized by active galactic nuclei (AGNs) and composite regions. We present a simple reprojection of the 3D line ratio space composed of [N II] λ 6583/H α , [S II] λ 6716, 6731/H α , and [O III] λ 5007/H β , which reveals its model-constraining power and removes the ambiguity for the true composite objects. It highlights the discrepancy between many theoretical models and the data loci. With this reprojection, we can put strong constraints on the photoionization models and the secondary nitrogen abundance prescription. We find that a single nitrogen prescription cannot fit both the star-forming locus and AGN locus simultaneously, with the latter requiring higher N/O ratios. The true composite regions stand separately from both models. We can compute the fractional AGN contributions for the composite regions, and define demarcations with specific upper limits on contamination from AGN or star formation. When the discrepancy about nitrogen prescriptions gets resolved in the future, it would also be possible to make robust metallicity measurements for composite regions and AGNs.

Key words: galaxies: active – galaxies: nuclei – galaxies: star formation.

1 INTRODUCTION

The widely used optical diagnostic diagrams (hereafter BPT diagrams) originally proposed by Baldwin, Phillips & Terlevich (1981) and refined by Veilleux & Osterbrock (1987) are very useful for distinguishing among different ionization mechanisms in galaxies. The merits of these diagnostics are that only ratios of strong optical emission lines are involved (i.e. [S II] $\lambda\lambda$ 6716, 6731/H α . [N II] λ 6583/H α , [O I] λ 6300/H α , and [O III] λ 5007/H β), and they are insensitive to dust extinction. Various demarcations have been proposed based on these diagrams over the years, which add to our understanding of the variety and complexity of the ionization mechanisms in the universe (Veilleux & Osterbrock 1987; Ho, Filippenko & Sargent 1997; Kewley et al. 2001; Kauffmann et al. 2003; Kewley et al. 2006; Ho 2008). Utilizing sophisticated photoionization codes like CLOUDY and MAPPINGS, people are able to match the distribution of the observed data in the BPT diagrams with the model predictions, and put constraints on the properties of ionized regions and/or sources (e.g. Dopita et al. 2000; Kewley & Dopita 2002; Groves, Dopita & Sutherland 2004b; Dopita et al. 2013).

However, there are several limitations associated with the usage of the BPT diagrams. First of all, there is often ambiguity between the classifications of different BPT diagrams. Spectra classified as AGN in one diagram could be classified as star forming (SF) in another diagram, and vice versa. These ambiguous cases are tricky to deal with and are often excluded in data analyses, leading to biased

The above issues could be largely resolved if one requires the consistency of the model predictions on more than two line ratios. A simple solution would be to compare models with data in a multidimensional line-ratio space. The traditional BPT diagrams can

or incomplete physical interpretations. In addition, photoionization predictions with significantly different model parameters (densities, abundance patterns, ionizing spectra, etc.) are able to cover the data space equally well (e.g. D'Agostino et al. 2019b). The lack of a merit function that requires consistency across multiple diagrams makes it difficult to evaluate the goodness of the fit. Partly this is because of the difficulty of visualizing model consistency across different diagnostic diagrams. Finally, apart from the degeneracy in the model parameters, there is also difficulty in the interpretation of the zone in-between the SF locus and AGN region in the BPT diagrams, which is commonly named as 'composite', 'transition', or 'ambiguous' region. The line ratios there could either be a result of mixing between different ionized clouds, e.g. SF and Seyfert, or SF and low-ionization nuclear emission-line regions (LINERS), due to insufficient spatial resolution, or it could be caused by the intrinsic variations in the physical parameters of Seyfert narrow-line regions or LINERs. It is hard to break this degeneracy with the traditional BPT diagrams. There are numerous efforts trying to decompose points in this zone into combinations of AGN and SF (e.g. Kewley et al. 2006). However, the semiarbitrary choices of the starting points in the AGN and SF zones in the BPT diagrams make those results unconvincing. In addition, this degeneracy makes it difficult to constrain metallicities or ionization parameters for data points in this zone and they are hardly attempted.

^{*} E-mail: xji243@uky.edu (XJ); rya225@g.uky.edu (RY)

be viewed as specific projections of this higher dimensional space. However, these straightforward projections are poorly positioned in revealing the discrepancy or consistency between models and data. A carefully chosen new projection from a more advantageous angle would be able to show those discrepancies more clearly. The angle of such a projection should be determined by the intrinsic shapes and orientations of the model surfaces embedded in higher dimensions. In this paper, we utilize the three traditional and most commonly available line ratios, [N II] λ 6583/H α , [S II] $\lambda\lambda$ 6716, 6731/H α , and [O III] λ 5007/H β , to construct a new set of re-projected optical diagnostic diagrams, which have the power to overcome all of the issues mentioned above.

The concept of requiring consistency across multiple line ratios have been applied in some cases, especially for the derivation of metallicities. Tremonti et al. (2004), for instance, used six strong optical emission lines and calculated the likelihood distribution of metallicity for their sample galaxies by comparing the measured line intensities with a large number of models. Similarly, by adopting the Bayesian inference, Blanc et al. (2015) developed the code IZI to consistently infer the metallicity and ionization parameter of H II regions given a set of observed strong emission lines and SF templates. In principle, these methods can also be applied to AGN narrow-line regions so long as the corresponding models are computed. However, without a viable visualization of the consistency between model and data in the higher dimensional space, models that are systematically offset from the data in the higher dimensions can still be used in these methods without users realizing it. In addition, these methods all rely on prior classifications into AGN and SF according to BPT diagrams, which would suffer the same issues mentioned above. Last but not least, all these methods would have difficulties when dealing with the 'composite regions'.

Vogt et al. (2014), on the other hand, directly looked into the distribution of galaxies in the 3D line-ratio space, and constructed a number of 3D diagnostic diagrams using different combinations of optical emission line ratios. By incorporating photoionization models for SF regions, they defined a series of 2D diagnostic diagrams projected from three dimensions. These diagrams are capable of showing the metallicity sequence of SF regions clearly, and their classification of different ionized regions is in good agreement with that based on the original [N II] BPT diagram. In this work, however, we take a further step by considering the photoionization models for AGN regions at the same time. Our preferred projection is defined by making not only the SF model surface, but also the AGN model surface compact. This definition puts a stronger constraint on the resulting viewing angles in three dimensions, and facilitates analyses on the composite regions. In addition, we carry out a detail comparison among models with different input parameters in the derived projection, in order to make a selfconsistent definition of theoretical demarcations for SF and AGN

In summary, while various attempts have been made in deriving the properties of ionized regions in high dimensions using photoionization models, seldom are the model parameters and assumptions carefully examined. Therefore, it is important to first verify the consistency between models and data in the higher dimensional space, which would then allow much more robust constraints on metallicity and ionization parameters. Our solution, a set of reprojected diagnostic diagrams obtained through rearranging the line ratios used in the original optical diagrams, not only enables strong constraints on the model assumptions, but also has the potential to make metallicity constraints much more robust and applicable to the composite regions.

In Section 2, we introduce the theoretical models and observational data we use. We show the comparison between this new diagram and the original BPT diagrams, and describe the procedure to construct this new diagram in Section 3. In Section 4, we discuss the implications and applications of the new diagram. Discussions about the robustness of our analyses, including the impact of sample selection on data distribution and the effect of time evolution on our photoionization models, are given in Section 5. We summarize our conclusions in Section 6. Throughout this paper, all wavelengths are given in air. For the three frequently used line ratios in this paper, i.e. [N II] $\lambda 6583/H \alpha$, [S II] $\lambda \lambda 6716$, $6731/H \alpha$, and [O III] $\lambda 5007/H \beta$, we denote their decadic logarithms as N2, S2, and R3, respectively.

2 MODELS AND DATA

The new optical diagnostic diagram we propose in this paper is determined by the overall shapes and orientations of the SF and AGN model grids in the line-ratio space, as this is a projection that compactifies both models simultaneously. A summary of the major parameters of the models is shown in Table 1. Note that this table only includes our fiducial and best-fitting models, and we will compare them to other models with different densities, ionizing SEDs, or nitrogen prescriptions in Section 4.1. The following is a description of the models and the data we use.

These models are generated using the photoionization code CLOUDY (Ferland et al. 2017). The input SEDs for the SF models are computed using the code STARBURST99 (Leitherer et al. 1999). We assume a Kroupa initial mass function (IMF), and a continuous star formation history over 4 Myr. SEDs with different stellar metallicities are computed, with $Z/Z_{\odot} = 0.05, 0.2, 0.4, 1$, and 2, respectively. The ionized cloud is set to have a hydrogen density of 14 cm⁻³ (at the ionizing face), which is the median value we obtain using the TEMDEN routine in PYRAF (De Robertis, Dufour & Hunt 1987; Shaw & Dufour 1995) for H II regions in MaNGA, assuming an electron temperature of 10⁴ K. The gas pressure is assumed to be constant throughout the cloud. We include dust grains with typical ISM abundance, which will also scale with the metallicity of the cloud. Metal depletion on to the dust grains is computed using the values given by Jenkins (1987) and Cowie & Songaila (1986). Finally, the cosmic-ray background of the local Universe is included.

To set up a grid, we vary the ionization parameter and the gasphase metallicity of the cloud. The ionization parameter is defined as $q \equiv \log U \equiv \log(\Phi_{\rm ion}/n_{\rm H}c)$, where $\Phi_{\rm ion}$ is the flux of ionizing photons at the illuminated surface of the cloud, and $n_{\rm H}$ is the volume density of hydrogen. The range we adopt for this parameter, q, is from -4.0 to -2.0. In the meanwhile, we vary the logarithmic gas-phase metallicity $[O/H] \equiv \log ((O/H)/(O/H)_{\odot})$ from -1.3 to 0.5(corresponding to 0.05–3.16 in linear space). For the SF models, we adopt a consistent stellar metallicity for the input SED as the gas metallicity when possible. Since our input SEDs only cover stellar metallicity up to Z/Z_{\odot} of 2, for all models with gas metallicities greater than this value, we use the SED with the highest metallicity. All elements except for helium, carbon, and nitrogen directly scale with the oxygen abundance. The solar abundance we use is taken from Grevesse et al. (2010), with $12 + \log (O/H)_{\odot} = 8.69$. The helium abundance is described by a primary production formula (Dopita et al. 2002), which gives the number density ratio of helium to hydrogen:

$$He/H = 0.0737 + 0.024 \times (O/H)/(O/H)_{\odot}.$$
 (1)

For carbon and nitrogen, contribution from the secondary production is important. How to quantitatively account for the yield

Table 1. Photoionization model sets.

Parameter	Values
	SF models
q	-4.0, -3.5, -3.0, -2.5, -2.0
[O/H]	-1.3, -0.7, -0.4, 0.0, 0.3, 0.5
$\log (n_{\rm H}/{\rm cm}^{-3})$	1.15
Ionizing SED	STARBURST99 models with $\log(Z/Z_{\odot}) = -1.3, -0.7, -0.4, 0.0, 0.3$
	Continuous SFH for 4 Myr
Nitrogen prescription	Dopita13 prescription
	AGN models
q	-4.0, -3.5, -3.0, -2.5, -2.0
[O/H]	-0.75, -0.5, -0.25, 0.0, 0.25, 0.5, 0.75
$\log (n_{\rm H}/{\rm cm}^{-3})$	2.0
Ionizing SED	Power-law SED with the power-law index $\alpha = -1.4$ (fiducial), -1.7 (preferred by data)
Nitrogen prescription	Groves04 prescription

of the secondary elements remains debated among literature. And there is evidence that the abundance of these elements could have dependence on the properties of galaxies, like the total stellar mass and the star formation efficiency (e.g. Belfiore et al. 2017; Schaefer et al. 2020). In this paper, we use the nitrogen prescription adopted by Dopita et al. (2013) for our SF models. Dopita et al. (2013) used a compilation of observed H II regions from van Zee et al. (1998). We refit this data using a simple quadratic function:

$$N/O = 0.0096 + 72 \times O/H + 1.46 \times 10^{4} \times (O/H)^{2}.$$
 (2)

When we adopt this nitrogen prescription, the carbon abundance is set to be always 1.03 dex larger than the nitrogen abundance, to be consistent with Dopita et al. (2013).

For the AGN models, however, we switch to the following nitrogen prescription described by Groves, Dopita & Sutherland (2004a) to fit the line ratios of AGN host galaxies:

$$N/O = 10^{-1.6} + 10^{2.33 + \log(O/H)}.$$
 (3)

Groves et al. (2004a) obtained this prescription by fitting a set of H II regions and nuclear starburst galaxies from Mouhcine & Contini (2002) and Kennicutt, Bresolin & Garnett (2003). When adopting this prescription, we set the carbon abundance to be 0.6 dex larger than the nitrogen abundance, to be consistent with the solar abundance. We will detail in Section 4.1 why we decide to apply different prescriptions to different photoionization models.

The form of the ionizing SED we choose for the AGN models is a broken power-law function adopted by Groves et al. (2004a). Our starting model assumes that $d\log F_{\nu}/d\log(\nu) = -1.4$, from $h\nu = 1$ Ryd to approximately 100 Ryd. This simple SED fits the observational data reasonably well in the original BPT diagrams. Ji et al. (2020) showed that this model provides a nice match to the upper boundary of the AGN region shown by their sample in the [S II] BPT diagram. Later in Section 4.1, we will use the SED with a power-law index of -1.7 in this intermediate energy range, as photoionization models with this softer SED match the majority of the AGN-ionized clouds better in the new diagram. The hydrogen density is set to be 100 cm^{-3} , which is the median value we found for the narrow-line regions (NLRs) in MaNGA (Ji et al. 2020). The ionization parameter for the ionized cloud varies from -4.0 to -2.0 and the logarithm of the metallicity varies from -0.75 to 0.75.

The observational data we use are obtained by MaNGA (Bundy et al. 2015; Yan et al. 2016b) and released as part of the 15th data release (DR15) of Sloan Digital Sky Survey (SDSS), which is identical to the 7th internal MaNGA Product Launch (MPL-7).

This data set includes spatially resolved spectroscopic data of 4621 unique galaxies. MaNGA is part of SDSS-IV (Blanton et al. 2017), and is the largest integral field spectroscopy (IFS) survey of galaxies to date. The observation is done using the 2.5-m Sloan Telescope (Gunn et al. 2006). MaNGA utilizes the BOSS spectrographs (Smee et al. 2013) and a number of fibre-bundle integral field units (Drory et al. 2015) to provide spectra with a wavelength coverage from 3622 to 10 354 Å, and a median spectral resolution of $R \sim 2000$. The data are reduced by the MaNGA Data Reduction Pipeline (Law et al. 2016), which produces datacubes with each spatial pixel (spaxel) being 0.5×0.5 arcsec². The relative flux calibration is accurate to 1.7 per cent between H α and H β (Yan et al. 2016a). The median full width at half-maximum of MaNGA's PSF is close to 2.5 arcsec (Law et al. 2015), which corresponds to a physical scale of \sim 1.5 kpc at a typical redshift of 0.03 (Wake et al. 2017). The emission-line measurements we present in this paper are performed by the MaNGA data analysis pipeline (DAP; Belfiore et al. 2019; Westfall et al. 2019), which uses the code ppxf (Cappellari & Emsellem 2004; Cappellari 2017) at its core.

Our sample consists of spaxels from the central regions of MaNGA galaxies, with $r/R_e < 0.3$, where r is the inclination-corrected galactic-centric distance of a given spaxel and R_e is the elliptical Petrosian effective semimajor axis measured in the SDSS r band of a given galaxy. We require that the signal-to-noise ratios (S/N) of the H α , H β , [O III] λ 5007, [NII] λ 6583, and [SII] $\lambda\lambda$ 6716, 6731 emission lines are all greater than 3. There is no cut in the ionization properties, so this sample include spaxels of SF-photoionized regions, AGN-photoionized regions, LINERs photoionized by AGNs, and low-ionization emission-line regions (LIERs) not photoionized by AGNs. The latter two categories are indistinguishable by strong-line ratios alone and we will group them as LI(N)ERs. We will discuss these LI(N)ER spaxels and their identification in Section 5.1.

3 NEW PROJECTION FOR THE OPTICAL DIAGNOSTIC DIAGRAM

3.1 Projections from the 3D line-ratio space

In Fig. 1, we plot our photoionization models for SF and AGN together with our sample in an animated 3D line-ratio space spanned by N2, S2, and R3. Each photoionization model manifests itself as a 2D continuous surface embedded in the 3D space, and each point on the model surface corresponds to a unique set of metallicity and ionization parameter. It is clear that there is a gap between the AGN

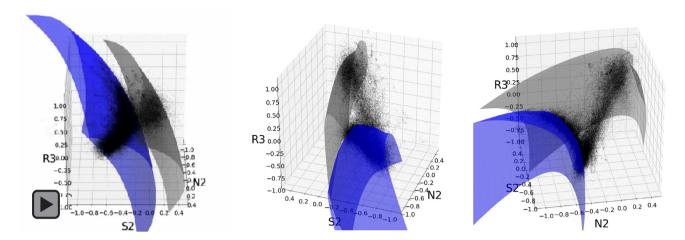


Figure 1. Left-hand panel: a 3D BPT diagram viewed at an elevation angle of 30° . A sample of MaNGA central spaxels with $r/R_e < 0.3$ are shown as black points. The SF and AGN models are plotted as the blue and the grey surfaces, respectively. The animated version of this figure is available in the online article. Middle and right panels: two still images from the animation, illustrating the gap between two model surfaces and how the data points are distributed relative to them.

model surface and the SF model surface, indicating that there is no degeneracy between these two ionization models, at least in the part of parameter space we care about. Line ratios of pure SF or pure AGN have to be located on or near these two disjoint model surfaces. Some data points are indeed there. However, some of the data points lie in the gap, separate from both model surfaces. They actually form a bridge connecting the two surfaces. The most natural explanation is that they are indeed 'composite' objects resulting from a combination of SF and AGN due to unresolved observations.

In Fig. 2, we show a few 2D projections of this 3D line-ratio space. Two of the projections correspond to the traditional [S Π]-based and [N Π]-based BPT diagrams, while the third one is a new projection we propose. The axes of the new projected 2D diagram (wihch we denote as P_1 and P_2) are made of linear combinations of the three line ratios from the two original diagrams, thus having the same virtue of using only strong lines and being insensitive to extinction. The advantage of the new diagram compared to the original ones is that it maximizes the separation between the SF model grids and the AGN model grids. In essence, we are choosing the projection that views both model surfaces roughly edge-on for the parts of parameter space that matter. The gap between them would become even emptier, if one further removes some models that are not found in the data, judging with the third dimension that is perpendicular to the two dimensions shown here. We will illustrate this below.

Furthermore, with the model surfaces appearing edge-on, it becomes easier to directly compare them with the observational data, in a way that maintains consistency across three line ratios. It not only enables us to better constrain the input parameters of the models, but also provides a way to investigate the contribution of different ionization sources to the observed line ratios. We will discuss these applications in Section 4.

3.2 Derivation of the axes of the new projection

Fig. 3 shows the data distribution and our models in the traditional BPT diagrams. From the 3D perspective, we could view the [N II] and [S II] diagrams as two different projections. The AGN model in the [S II] projection is closer to being 'edge-on' while it is nearly 'face-on' in the [N II] one. This is why the data distribution in the [S II] diagram is more useful for constraining the AGN SED (Ji et al. 2020). In contrast, the [N II] projection works better to isolate

pure H II regions because the SF surface becomes more compact at high metallicity in this diagram. It is natural to imagine that there is a projection that makes both surfaces edge-on, which would be more ideal to reveal the gap between model surfaces and to separate regions with different ionization mechanisms.

Here we show how to obtain the ideal projection by consecutive rotations in the 3D line-ratio space. To begin with, we note that for a small enough region on the model grid, the curvature of the model surface is negligible. If we have two planes with exactly zero curvatures, the procedure to find their edge-on projection can be summarized as follows:

- (1) Find where the two planes intersect. It will be a straight line in the 3D space.
 - (2) Find a plane which is perpendicular to this intersection line.
- (3) Find the two axes of this plane. We note that in this step, one has the freedom to change these axes under a rotation about the intersection line. An easy way to find a set of axes is to perform two consecutive rotations of the original axes. The direction of the intersection line determines a polar angle θ and an azimuthal angle ϕ . If we start with N2, S2, and R3 as x-, y-, and z-axes, a first rotation about the z-axis by an angle ϕ and a second rotation about the new y-axis by an angle θ would directly lead us to the plane (which is the final P_1 – P_2 plane).

Going back to our model surfaces, they are curved surfaces in the 3D diagram. But since we only care about the part of parameter space where our sample is located, we can pick out certain patches of the model surfaces, which will be small enough to be approximated by planes. In Fig. 3, we choose one quadrangle patch from each photoionization model grid. The patch of SF grid covers the part of the parameter space satisfying $0.0 < [{\rm O/H}] < 0.3$ and -2.5 < q < -3.0, and the patch of the AGN grid covers the parameter space satisfying $0.0 < [{\rm O/H}] < 0.25$ and -3.0 < q < -3.5. These parts of models are chosen as they represent the line ratios of the the most densely populated areas by the data in the [N II] and [S II] BPT diagrams. The curvature of each patch is relatively small, and we can construct two planes by selecting three points from each of the quadrangle. Using these two planes, we are able to find the desired final plane of projection.

Fig. 4 shows how we construct the final plane in the 3D lineratio space. The polar angle θ (from the R3 axis) is 36°, and the

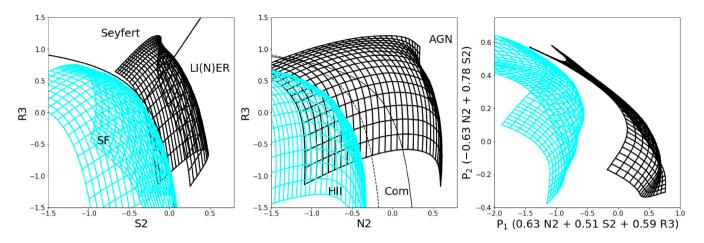


Figure 2. Comparison of photoionization models in three optical diagnostic diagrams. AGN model grids (black) and SF model grids (cyan) displayed in the diagrams have been smoothed through interpolation. Left-hand panel: the [S II]-based BPT diagram; middle panel: the [N II]-based BPT diagram; right panel: the new diagram proposed by this paper.

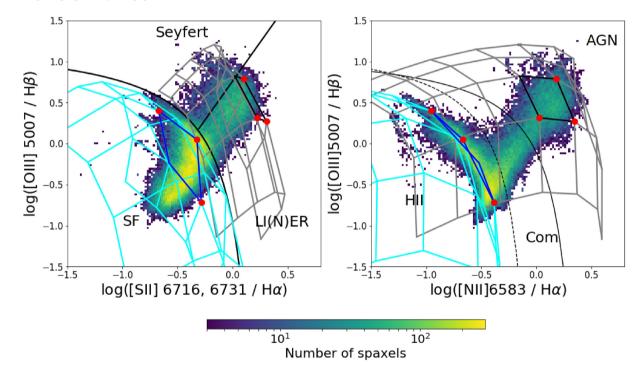


Figure 3. Photoionzation models and observational data viewed in the [S II]- and [N II]-based BPT diagrams. The solid black curves are Kewley et al. (2001) extreme starburst lines. The dashed black curve in the right panel is the Kauffmann et al. (2003) demarcation that separates H II regions from AGNs. The solid straight black line in the left panel is the Kewley et al. (2006) line that separates Seyferts from LI(N)ERs. The models used here are described in Section 2. The parts of the model grids that are highlighted in black (blue) is the patch we select to represent the AGN (SF) model surface. Red points are chosen to construct planes in the 3D diagram to find the angle from which both model surfaces are viewed nearly 'edge-on'.

azimuthal angle ϕ (defined as a counter-clockwise angle from the positive direction of the N2 axis) is 39° .

With the rotation angles determined, we calculate the expressions for the axes of this reprojected plane, which are

$$P_1 = 0.63 \text{ N}2 + 0.51 \text{ S}2 + 0.59 \text{ R}3 \tag{4}$$

and

$$P_2 = -0.63 \text{ N}2 + 0.78 \text{ S}2. \tag{5}$$

In the left panel of Fig. 5, we plot our data and models in this new diagnostic diagram. We cut the model grids to keep only those models

that cover a similar range on the P_3 axis (which is perpendicular to the P_1 – P_2 plane in the 3D space) as our data. The information from this third dimension further shrinks the spread of the models in the P_1 – P_2 plane. The expression for P_3 is

$$P_3 = -0.46 \text{ N}2 - 0.37 \text{ S}2 + 0.81 \text{ R}3.$$
 (6)

We note that one has the freedom to make a translation to the origin along the P_3 axis, which is equivalent to adding a constant term to the above equation. The sample distribution on the P_1 – P_3 plane is shown in the right panel of Fig. 5. Of our data, 98 per cent lie in the

5754

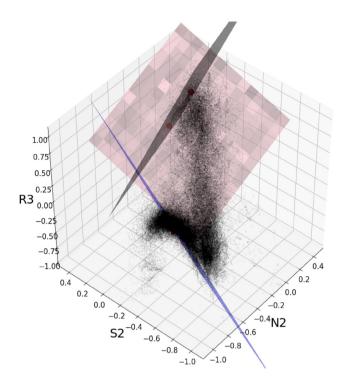


Figure 4. The final plane (pink) constructed from the two model patches (in grey and cyan, respectively). The black (blue) plane is the representative plane for the AGN (SF) model constructed through the red points, and is perpendicular to the final plane. The black data points are our sample spaxels.

range $-0.23 < P_3 < 0.98$. Thus, we keep only the parts of the model grids with P_3 within this range.

The final photoionization models appear quite compact in this P_1-P_2 projection. This leads to strong constraints on the model parameters, as the data points for pure AGN-ionized or SF-ionized clouds should trace the model surfaces. The data show a curved locus on the left-hand side, with P_1 between -1 and -0.5, which is traced nicely by our fiducial SF model. To the right of it, we see a nearly horizontal extension, with a concentration of points around $P_1 \sim 0.5$. This blob on the right contains the Seyferts and LI(N)ERs identified by the original BPT diagram. Our fiducial AGN model roughly traces the envelope of this blob rather than going through the densest part of it. This could mean that our fiducial model is not a good description for a typical AGN-photoionized cloud. We will investigate this in the next section. The stretch of points between the SF locus and the AGN blob are far away from both models. As we will demonstrate in the next section, they cannot be explained by varying the parameters of the SF model or the AGN model. The most natural explanation is that they are composites of SF and AGN, due to insufficient spatial resolution of our observations or spatial projection effects in the target galaxy.

Another interesting fact of this new projections is that the coefficients of N2 and S2 for the P_2 axis are very close to be opposites, making P_2 close to the S2N2 ratio, which is a good tracer for metallicity (Dopita et al. 2016). Thus, we expect the constant metallicity lines in this diagram to have roughly constant P_2 values. On the other hand, for the part of the model grids we kept, the constant ionization parameter lines have roughly constant P_3 values, as is shown in the right panel of Fig. 5. The outcome of these facts combined is that each iso-metallicity surface almost appears as a line parallel to the P_1 axis in the regime where our data points are populated, which facilitates the decomposition of the different

ionizing components for our sample, as we will discuss in the next section.

We list in the following several potential applications of this new diagram:

- (1) Separate the pure SF regions, pure AGN regions, and composite regions with no ambiguity. Since this projection already implicitly enforces consistency between the [N II] and [S II] BPT diagrams, and significantly reduces the projected areas of theoretical model grids, data points of pure SF and pure AGN should distribute about the projected models, and those of the composite regions should lie in the gap between them. One potential uncertainty is that there could be variations of the model positions due to different sets of model parameters. This point is further investigated in Section 4.1. The demarcations for different ionized regions in the new diagram based on the decomposition of line ratios are given in Section 4.3.
- (2) Constrain the photoionization model parameters, like the secondary nitrogen abundance and the shape of the ionizing SED. The data distribution spans a narrow range, less than 0.8 dex, along the P_2 axis, which is very sensitive to the amount of nitrogen included. Changing the nitrogen prescription should result in noticeable shift of the model relative to the data in the vertical direction, allowing one to put strong constraints on the nitrogen prescriptions. On the other hand, changing the hardness of the input SED will systematically lower/increase all line ratios. Since the vertical axis tends to cancel out the changes in line ratios, the model will move in the horizontal direction.
- (3) Estimate metallicity for composite regions. For the composite regions with line ratios sitting between pure-AGN and pure-SF model surfaces, it is potentially possible to construct iso-metallicity lines between the two edge-on models and calculate the metallicity for each point. However, there is still certain difficulty due to the nitrogen prescription inconsistency between the best-fitting AGN and SF models. We will illustrate this in Section 4.2.
- (4) Calculate the contributions of different ionizing sources, given the corresponding models. Since regions ionized by a single source should ideally have a very narrow distribution about the corresponding model surface, we are able to set up a parameterization so that every point lying between two different ionizing models can be described by the relative contributions from these sources. These results can be used to create demarcation lines with a specific threshold on the contamination from the other sources. But one should be cautious about the assumptions made before decomposing the sources.

We will give some preliminary examples of these applications in the next section.

4 IMPLICATIONS AND APPLICATIONS OF THE NEW DIAGRAM

In this section we discuss both the implications and applications of the new projection for the optical BPT diagnostics.

4.1 Constraints on model parameters

With our new P_1 – P_2 projection, we consider that the correct SF model should go through the densest part of the SF locus on the left and the correct AGN model should go through the densest part of the AGN blob on the right. There are two reasons. First, the data should show scatters around the model due to intrinsic scatter of the physical parameters and observational uncertainty on the line ratios. Secondly, for the composite objects, the most natural line ratio

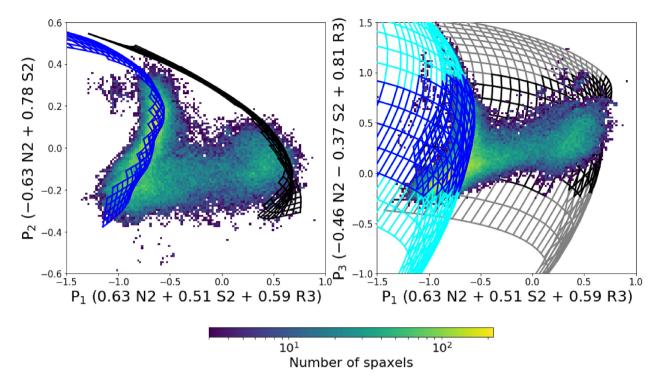


Figure 5. Left-hand panel: final plane of projection. The photoionzation models are first interpolated and then cut in the P_3 (perpendicular) direction so that they cover the middle 98 per cent of the data points. Right-hand panel: P_1 – P_3 plane. Parts of the model grids (in cyan and grey) are cut off according to the distribution of the data points along the P_3 axis. The 2D number density distributions of the our sample spaxels are displayed in both panels.

distribution of them should pile up at the two end points where one component completely dominates over the other. Any concentration of data in-between the end points would require an explanation of why that particular ratio of combination is preferred. Therefore, we think the densest regions of the SF locus and AGN blob should match the pure SF and AGN models, respectively.

Thus, we first examine how the parameters of the photoionization model can impact the position of the models in this new diagram. In particular, we need to find what changes to the AGN model parameter could lead to a better fit to the densest part of the AGN blob. We will see that the data distribution in this new projection can place stringent constraints on the models. We consider three important inputs to the photoionization models: the hydrogen density of the cloud, the secondary nitrogen abundance (since we are using the [N II] emission line), and the shape of the input ionizing SED.

Among the three inputs, the hydrogen density is well constrained by the electron density derived using the [S II]λ6716/[S II]λ6731 ratio of our sample. The PYRAF calculation gives a median value of 100 cm⁻³ for the AGN spaxels and 14 cm⁻³ for the SF spaxels in MaNGA. This is in good agreement with the result we obtain by directly comparing the [S II] doublet ratios produced by CLOUDY models to the median ratio in the data. Thus, we do not need to change this assumption. Regardless, we illustrate here how density variation will impact the models in our new projection, which demonstrates the power of looking from the right angle. Fig. 6 shows this density dependence of the models. At low metallicities and when density is less than 100 cm⁻³ for SF or 1000 cm⁻³ for AGN, the models are not sensitive to the densities. But at high metallicities, both the SF and AGN models shift to the right with an increasing density. At even higher density, the AGN model grid moves downward (at high metallicity) and even to the lower left (at low metallicity). The movement of the model surface is mostly due to the temperature

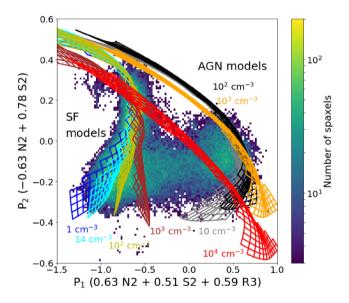


Figure 6. SF and AGN models with various hydrogen densities in our new projected diagram. Our fiducial SF model has a density of 14 cm^{-3} , and our fiducial AGN model has a density of 100 cm^{-3} . For SF models, we compute four densities from 1 to 10^3 cm^{-3} , and for AGN models, we compute four densities from $10 \text{ to } 10^4 \text{ cm}^{-3}$. The density values are shown by the side of the corresponding models. All the models are cut so that they enclose the middle 98 per cent of the data along the P_3 axis.

dependence of the collisional excitation rates for [S II] and [N II]. The change in direction above a density of $1000~\rm cm^{-3}$ is due to the collisional de-excitation of [S II]. These movements are very significant compared to the data concentration. In the traditional BPT diagrams (see e.g. fig. 11 of Ji et al. 2020), the role of density is

much less obvious because it is all hidden by projection effects due to the non-optimal viewing angles.

In comparison to the density, the nitrogen abundance prescription is more difficult to measure, which results in significant disagreements in the literature. Though Schaefer et al. (2020) derived a median N/O ratio versus metallicity relation for H II regions from MaNGA, they also find that this relation depends on the total stellar mass and the local star formation efficiency of the host galaxies. It is possible that AGN-ionized regions (and/or LIERs) have different nitrogen enhancement from that of the SF regions, as they are more likely to be found in more massive and evolved galaxies. In order to constrain this parameter from the side of theoretical models, we need to understand how the model grids will move as a whole when the nitrogen prescription is changed. Qualitatively speaking, as the nitrogen abundance at a given metallicity increases, the N2 ratio should increase as well. This makes the model grid move in the lower right direction, as indicated in the left panel of Fig. 7. The exact behaviour of the model grid might well differ for parts with different metallicities, as can be seen in the case of the AGN models, where the model also seems to rotate a little bit after we switch its nitrogen prescription. The two sets of AGN models we plot in this figure are generated with nitrogen prescriptions adopted by Groves et al. (2004a) and Dopita et al. (2013) (we denote them as Dopita13 prescription and Groves04 prescription hereafter), respectively. The corresponding log(N/O) versus 12 + log(O/H) relations are shown in Fig. 8, together with some other relations we found in the literature (cf. Vila Costas & Edmunds 1993; Storchi-Bergmann et al. 1998; Nicholls et al. 2017; Schaefer et al. 2020). We choose these two relations because they yield the highest and the lowest N/O ratios for all metallicities. The separation between the two sets of model grids basically shows the largest difference in line ratios induced by the variation in the N/O ratio. For SF regions, we add another model with the nitrogen prescription derived by Schaefer et al. (2020, hereafter Schaefer20 prescription), which gives the second lowest N/O ratios for most metallicities as shown in Fig. 8. To fit the SF locus traced by the densest part of the data distribution, the lowest N/O ratio prescribed by Dopita et al. (2013) is needed. A slightly elevated N/O versus [O/H] relation like the Schaefer20 prescription would be enough to drive the model away from the centre of the SF locus. As a consequence, the SF locus in this diagram puts a strong constraint on the nitrogen prescription for our SF models. On the other hand, an AGN model with the Dopita13 prescription would lie too high to match the AGN blob in our new diagram centred around $P_1 \sim 0.4$ and $P_2 \sim -0.1$.

As a result, it seems that there is a discrepancy between the best-fitting nitrogen prescriptions indicated by the SF and AGN models. Raising the hydrogen density of the AGN model to much higher than $1000~\rm cm^{-3}$ could lower its vertical position. But this is not a viable solution as it would contradict the measured [S II] doublet ratio.

Another model input we can vary is the SED of the ionizing spectra. The right panel of Fig. 7 shows that by softening the input SED of the AGN model, we can move the model grid to the left and slightly downward. An AGN model with a softer SED and the Dopita13 nitrogen prescription could match the upper part of the AGN blob, but would require unrealistically high metallicity ([O/H] >0.75) to cover the bottom part of it. Therefore, varying the SED would not solve the discrepancy in the nitrogen prescription between SF and AGN models.

In fact, if we use the density distribution of the AGN blob as a metric, to make the AGN model go through the densest part of it requires a Groves04 nitrogen prescription and a power-law index of \sim 1.7, as shown in the right panel of Fig. 7.

If we consider adjusting the nitrogen prescription of the SF model instead, the SF model with a Groves04 nitrogen prescription would fall too far to the lower right-hand side of the SF locus. It seems even harder to move it up in the upper left-hand direction by changing the density or the shape of the ionizing SED. In the right panel of Fig. 7, we show several SF models with different input SEDs corresponding to different ages for a starburst with constant star formation rate. The models do not differ in their vertical positions at all. According to D'Agostino et al. (2019b), possible variations in the ionizing radiation field of the H II regions from different population synthesis models only introduce an average systematic difference of \sim 0.1 dex in the N2 and R3 line ratios. Thus, varying the input SEDs for the SF models also could not resolve this discrepancy.

To summarize, solely judging from the distribution of the data in our new diagram, the best-fitting SF model has a Dopita13 nitrogen prescription and a continuous SFH no younger than 4 Myr, while the best-fitting AGN model has a Groves04 nitrogen prescription and a power-law index of -1.7. This discrepancy in the nitrogen prescription cannot be easily solved by adjusting other model parameters. This discrepancy suggests some assumptions we make in modelling the ionized cloud must be incorrect, or there are some difference in the physical environments around SF and AGN that we have not yet considered in our modeling. We will further discuss this puzzle and its consequence in the next section.

4.2 AGN/SF composites and the iso-metallicity mixing sequence

With the above exploration of how AGN and SF models depend on the model parameters and assumptions, we found that none of those models could completely cover the zone between the SF locus and the AGN blob. Thus, the simplest explanation is that these data points are from truly composite regions made up by combining SF-ionized clouds with AGN/LIER-ionized clouds.

We would like to clarify what we consider as the correct physical picture for a composite region. We consider that a composite region is composed of spatially close but physically separate clouds with different ionizing sources, some are ionized by massive young stars (SF), and others by AGN or whatever source is responsible for LIERs or diffuse ionized gas. The spectra of these clouds are combined together due to insufficient spatial resolution of the observations or projection effects in the target galaxy, leading to a set of line ratios that is intermediate between SF and AGN/LIERs.

We consider that it is unlikely for the spectra of composite regions to come from clouds that are ionized by a composite SED made up by comparable fractions of AGN photons and SF photons, as it requires fine tuning to make them to be on the same order of magnitude. Given that radiation flux falls with distance squared, it is more likely that they differ by orders of magnitude for a random location in the galaxy. Thus, a single cloud is nearly always dominated by one of the ionizing sources. In galaxies with significant shock ionization, the same principle should apply that each cloud is dominated by one of the ionization mechanisms rather than balanced by comparable contributions from more than one.

Under this physical picture, it is reasonable to assume the AGN/LIER-ionized clouds and SF-ionized clouds that make up a composite region are spatially close (on scales smaller than the PSF of the observations) and have roughly the same metallicities and abundance pattern.

Given this picture and this assumption, it is now feasible to decompose the composite regions into AGN and SF components, and at the same time, constrain their metallicities. Our new projection of the line-ratio space makes these constraints quite apparent and much

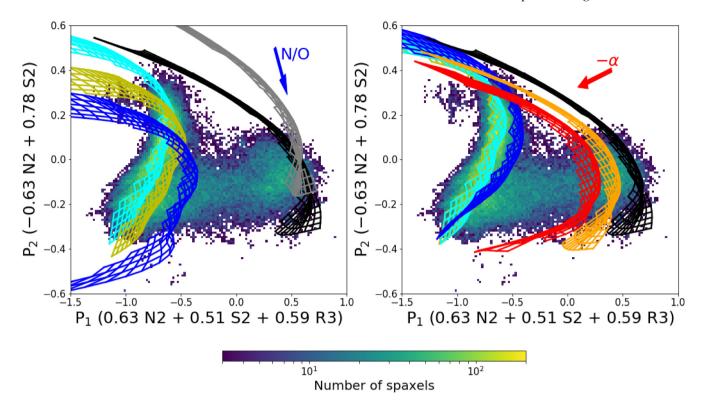


Figure 7. Photoionization model grids and the distribution of MaNGA central spaxels in the new diagram. The colour-coding of the data indicates the density of spaxels in the diagram. Left panel: the SF model with the Dopita13 (Groves04) nitrogen prescription is shown in cyan (blue), and the AGN model with the Dopita13 (Groves04) nitrogen prescription is shown in grey (black). Besides these two extreme prescriptions, the Scheafer20 prescription is used to compute the yellow model grid for SF regions. The blue arrow roughly indicates how the model grids would move if the overall N/O ratio is increased. Right-hand panel: AGN models with a Groves04 nitrogen prescription and input SEDs of which the power-law indices are -1.4, -1.7, and -2.0 are shown in black, orange, and red, respectively. The red arrow roughly indicate in what direction the AGN model grid would move if the power-law index of the input SED is decreased. SF models generated by SEDs of continuous star formation history but different ages are plotted as well. The SF ages for the blue, cyan, and yellow models are 0.01, 4, and 8 Myr, respectively. All the models are cut so that they enclose the middle 98 per cent of the data along the P_3 axis.

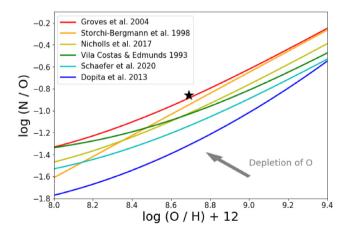


Figure 8. Various nitrogen prescriptions in literature. The Groves04 and the Dopita13 are the two prescriptions used in this paper. The solar abundance is indicated by the black star. The arrow shows the effect of the depletion of oxygen on to dust grains by 0.22 dex. By default, CLOUDY assumes that oxygen depletes by 40 per cent or approximately 0.22 dex, while nitrogen does not deplete.

less ambiguous than decompositions based on the original [N $\rm II$]- and [S $\rm II$]-based diagrams. As long as we can find consistent abundance prescriptions between AGN and SF regions, we can connect those AGN models with SF models at the same metallicity and define

iso-metallicity mixing sequences. The fact that the ionization parameter is nearly hidden in this P_1 – P_2 projection means the decomposition is nearly independent of the ionization parameters of the AGN and SF regions.

However, as we discussed in Section 4.1, we cannot find a single nitrogen prescription to fit both the SF locus and the AGN blob. Without such a consistent abundance pattern prescription, we cannot carry out a self-consistent decomposition and metallicity determination for composites. Regardless, we show here how the iso-metallicity sequences would look like in our new projection, and how they compare to the distribution of composite regions, which will further demonstrate the tension in the nitrogen prescription between the SF and AGN models.

We begin by setting up a series of parameterized iso-metallicity surfaces (or line bundles) connecting our best-fitting AGN model and SF model in the 3D diagram. Since we adopt different nitrogen prescriptions for the AGN and SF models, the 'iso-metallicity' here really means iso-[O/H], but not iso-(N/O). We show below an example of how we parametrize a mixing line between a predetermined start point on the AGN surface and an end point on the SF surface. The start point is assumed to have line ratios of a pure AGN region, which we denote as N2_{AGN}, S2_{AGN}, and R3_{AGN}. Similarly, the end point corresponds to a pure SF region with line ratios N2_{SF}, S2_{SF}, and R3_{SF}. Now if we observe the two regions together, the resulting line ratios will change, according to the relative strength of the emission lines from different regions. Take N2 as an example, the

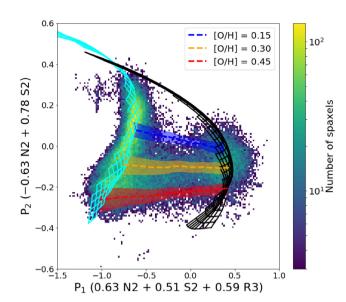


Figure 9. Examples of iso-metallicity line bundles constructed from the photoionization models. The dashed lines are the median trends of line bundles with different metallicities. The coloured shaded regions represent the projected areas within one stander deviation from the median along the vertical axis.

observed value N2_{obs} can be calculated using the following equation:

$$N2_{\text{obs}} = \log(\frac{([N \text{ II}]/H \alpha)_{\text{AGN}} + i ([N \text{ II}]/H \alpha)_{\text{SF}}}{1 + i}), \tag{7}$$

where i is defined as the ratio of the H α flux contributed by the SF region to that contributed by the AGN region, i.e. $i \equiv H \alpha_{SF}/H \alpha_{AGN}$. S2_{obs} can be computed using a similar equation. Whereas the equation of R3_{obs} involves the SF-to-AGN ratio for H β flux, which we denote as j. Note j is not necessarily the same as i because the SF-ionized clouds and the AGN-ionized clouds could have very different dust extinction, although they are spatially close. To relate i and j, we need to assume a ratio between the Balmer decrement for the AGN clouds (b_{AGN}) and that for the SF clouds (b_{SF}) , which we denote as $\Theta \equiv b_{AGN}/b_{SF} = j/i$. We can make a simplification by choosing a constant value for Θ . By taking the ratio of the median Balmer decrement among the AGN regions to that among the H II regions in our sample, we obtain a representative value of 1.35 for Θ . We will use this value in the following analysis, i.e. j = 1.35i. The difference in Θ would lead to a different decomposition result but will not shift the iso-metallicity sequence significantly.

As a result, there is effectively one parameter for constructing a single mixing sequence that connects a start model point and an end model point. We can parametrize this sequence by changing $\log(i)$ from $-\infty$ to ∞ . By varying the ionization parameter of the start model point and the end model point at fixed metallicity, we can generate a bundle of such lines describing the overall trend of an iso-metallicity mixing sequence.

Fig. 9 shows three examples of the projected iso-metallicity surfaces (or line bundles) with [O/H] = 0.15, 0.30, and 0.45. The spread of each line bundle is caused by the variation in the ionization parameter. To fully utilize the information contained in the 3D distribution of the data, we re-cut the model grids so that they only enclose the middle 98 per cent composite spaxels (i.e. excluding the upper part of the SF locus with roughly $P_3 > 0.4$ as shown in the right panel of Fig. 5). While the median- and high-metallicity line bundles appear nearly horizontal, the low-metallicity line bundle

show a slightly negative slope. It appears that the composite spaxels in MaNGA follow a similar trend as our theoretical iso-[O/H] sequences. However, we note that the two ends of the composite sequence in the data are not necessarily having the same metallicity as they could come from different galaxies.

A sanity check for whether our models produce a sensible composite sequence is to compare the derived iso-metallicity lines with the overall shape of the composite zone. In Fig. 10, we perform a more careful comparison by binning our data according to the total stellar masses of their host galaxies. This would help identify iso-metallicity composite sequences in the data by making sure the two ends of them are from similar types of galaxies. The stellar masses are drawn from the MaNGA Firefly value-added catalog (Goddard et al. 2017). Within each stellar mass bin, we expect both [O/H] and N/O are roughly fixed with small scatters, and the distribution of the spaxels in the composite zone provides the correct slope for an iso-metallicity line bundle. Here we are assuming that the mass-metallicity relations for SF galaxies and AGN host galaxies are identical (for a potential counterexample, however, see Thomas et al. 2019). The theoretical iso-[O/H] line bundles produced by our best-fitting models all exhibit shallower slopes than those indicated by the median data trends in five different mass bins, which include most spaxels in the composite zone. To steepen the slopes of these iso-metallicity lines by changing the abundance pattern of the photoionization models, one could either lower the N/O versus [O/H] relation for the AGN model, or elevate this relation for the SF model. In the middle and right panels of Fig. 10, we plot the iso-metallicity lines using SF and AGN models with consistent nitrogen prescriptions, Dopita 13 for the middle panel and Groves04 for the right panel. We can see that lowering the N/O ratio raises the position of the AGN model in the diagram, and the resulting slopes of the iso-metallicity lines are now steeper compared with the data. A noticeable problem with lowering the N/O ratio of the AGN model is that the composite zone cannot be fully covered. Since the SF locus only reaches a highest [O/H] of 0.5 according to our SF model, if we only consider the corresponding part of the AGN model (which is coloured in black), even the iso-metallicity line bundle with the highest [O/H] could not overlap with the lower part of the mixing branch. On the other hand, if we raise the N/O ratio of the SF model instead, as is shown in the right panel of Fig. 10, the slopes indicated by our models for the composite zone are again too steep, and the SF model does not match the SF locus in the data.

To summarize, the nitrogen prescription for the SF model is strongly constrained by this SF locus, as we have already seen in Section 4.1. However, in order to cover the AGN blob in the data, the AGN model needs a prescription with higher N/O ratios, which results in an inconsistency in the abundance pattern to explain the composite galaxies. Furthermore, forcing the SF and AGN models to have the same N/O prescriptions makes the iso-metallicity mixing-sequences to be too steep compared to the trends of composite galaxies with fixed stellar masses. It could be due to incorrect assumptions or missing physics in our photoionization models, and we will investigate this issue with more sophisticated photoionization models in future work. Regardless, this illustrates nicely the power of our new projection of the line ratio space.

4.3 Decomposition of the ionized regions

In the previous section, we introduce a way to construct isometallicity line bundles (or surfaces) between the AGN and SF model grids, assuming the Balmer decrement ratio, Θ , is fixed. As a result, there is effectively one free parameter that determines the position

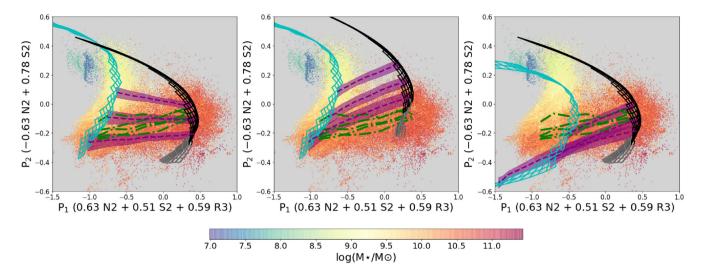


Figure 10. Comparisons of iso-metallicity line bundles based on different combinations of SF (cyan) and AGN (black and grey) models. The three purple bands correspond to three iso-metallicity line bundles in with [O/H] = 0.15, 0.30, and 0.45. All spaxels are colour-coded according to the total stellar masses of their host galaxies. The dot-dashed green lines trace the median trends of five stellar mass bins, from $log(M_{\star}/M_{\odot}) = 9.75$ to 11.00 with an interval of 0.25 dex. In the left panel, an SF model with the Dopita13 nitrogen prescription and an AGN model with the Groves04 nitrogen prescription are shown; in the middle panel, both models are generated using the Dopita13 prescription; in the right panel, both models are generated using the Groves04 prescription. The parts of the AGN models with [O/H] > 0.5 are coloured in grey.

of a point on a given iso-metallicity line, which we can choose to be the ratio of the H α flux from the SF-ionized clouds to that from the AGN-ionized clouds, i. Alternatively, we can use i to derive another physical parametrization. The fractional contribution to the total H α flux by the AGN-ionized clouds is H $\alpha_{\rm AGN}$ /H $\alpha_{\rm total} = 1/(1+i)$, which we denote as $f_{\rm AGN}$ hereafter. This parameter tells us how significant the contribution from AGN-ionized clouds is in a composite spectrum.

The decomposition process is summarized as the following. First, using a pair of pre-determined SF and AGN models, we construct all possible iso-metallicity lines (for all ionization parameter combinations), with the i parameter ranging from 10^{-3} to 10^3 equally spaced in the logarithmic space. For each data point, we compute the probability distribution of the models using Bayesian statistics, assuming a flat prior (i.e. equal prior probability for all model grid points). We compute the likelihood of each data spaxel, given each model point in the original 3D parameter space (N2, S2, R3) with the uncertainty in these three line ratios (in log). We do the calculation in the original 3D space as there is less covariance among them than there is among P_1 , P_2 , and P_3 . Given the resulting probability distribution of the models, we compute the expected value for $\log i$ and convert it to $f_{\rm AGN}$ for each spaxel.

The choice of photoionization models will have a non-negligible influence on the decomposition result. Since we currently cannot settle the nitrogen discrepancy while matching the models to the data, there should be no self-consistent way to decompose the data either. Still we would like to show in the following a demonstration of the decomposition using our best-fitting SF and AGN models (the set shown in the left panel of Fig. 10). Given that these models can well describe the majority of the AGN and SF spaxels, they should not be far from the correct models. But the correct models could have different slopes for the iso-metallicity lines. Judging from the left panel of Fig. 10, the bias on $f_{\rm AGN}$ due to different slopes of the iso-metallicity lines should not be too significant.

In Fig. 11, we plot the derived f_{AGN} as a function of the position along the P_1 axis. Due to the spread induced by the variation in the metallicity and the ionization parameter, the relation has relatively

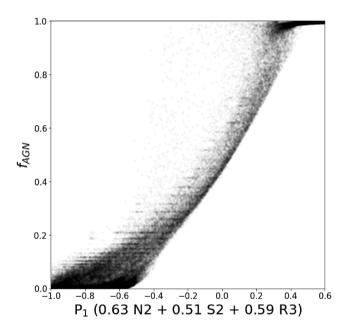


Figure 11. The derived f_{AGN} as a function of the position of the spaxel along the P_1 axis.

large uncertainties at both small P_1 values and large P_1 values. Part of the data points also seem to cluster at discrete values of $f_{\rm AGN}$, which might be due to our insufficient sampling of log i. Despite the spread, we can see that $f_{\rm AGN}$ increases more rapidly as P_1 gets larger, but quickly plateaus after P_1 reaches 0.3, which is roughly the horizontal position of the lower part of the AGN model. The correlation between $f_{\rm AGN}$ and P_2 , or that between $f_{\rm AGN}$ and P_3 , is much weaker.

If the Balmer decrement ratio Θ is changed, the horizontal position of each parametrized point would change accordingly. Increasing Θ would shift the model points to smaller P_1 values, hence increasing the overall fitted $f_{\rm AGN}$ of our sample. For example, if the Θ in our

setting is doubled (i.e. $\Theta=2.70$ in this case), the fractional change in the derived $f_{\rm AGN}$ of our sample has a median value of 12 per cent, and the the values for the 16th and 84th percentile of the data are 2 and 22 per cent, respectively. One may wonder if we can use the observed Balmer decrement to put constraints on Θ . This is not sufficient because the observed Balmer decrement also depends on the Balmer decrement in the SF-ionized clouds, $b_{\rm SF}$. Therefore, adding this additional constraint does not remove the need to make one assumption. In principle, if two additional dust-sensitive line ratios were added as extra constraints, it would be possible to solve for both $b_{\rm SF}$ and Θ .

Fig. 12 shows a map of all MaNGA central spaxels colour-coded by the average $f_{\rm AGN}$ of each line-ratio bin. With this map, we can generate demarcation lines that correspond to a constant fractional contribution to the total H α flux by AGN-ionized clouds. The dashed blue line corresponds to $f_{\rm AGN} \sim 0.10$, which means the spaxels to their left have a maximum contamination of 10 per cent from AGN-ionized clouds. Similarly, the spaxels to the right of the dot–dashed black line have a maximum contamination of 10 per cent from SF-ionized clouds. We can fit P_1 as a quadratic function of P_2 for these demarcation lines. The functional form of the demarcation line that identifies SF-dominated regions is

$$P_1 < -1.57P_2^2 + 0.53P_2 - 0.48 \text{ (for } -0.4 \lesssim P_2 \lesssim 0.1),$$
 (8)

and the functional form of the demarcation line that identifies AGN-dominated regions can be written as

$$P_1 > -4.74P_2^2 - 1.10P_2 + 0.27$$
 (for $-0.4 \le P_2 \le 0.1$). (9)

These demarcations are largely determined by the spaxels inside the composite zone, and therefore should only be applied when $-0.4 \lesssim P_2 \lesssim 0.1$. From the map (and also Fig. 11), we can see that the AGN fraction $f_{\rm AGN}$ drops more rapidly near the start points of the AGN model. The demarcation lines for AGN regions are largely determined by the location of the best-fitting AGN model grid. In the right panel of Fig. 12, we compare these two demarcations with the density distribution of our sample. While SF demarcation line clearly traces the outskirt of the SF locus, the AGN demarcation line is close to the centre of the AGN blob.

Since we are using the densest area at the tip of the mixing zone to constrain the AGN model, the location might well change for different samples. For example, a sample with systematically higher Eddington ratios might favour a photoionization model generated by an SED with larger optical to X-ray ratio, thus change the locations of our demarcation lines based on the decomposition (Lusso et al. 2010). We also note that potential contaminations from LI(N)ERs that have similar line ratios to AGN-ionized clouds but are not truly ionized by AGNs might hamper the determination of the best-fitting AGN model. However, if these pure LI(N)ER spaxels cluster at the same location of the true AGN-ionized clouds in the diagram, the decomposition still applies as the positions of the start points remain the same. The only difference is that instead of obtaining f_{AGN} , one obtains $f_{LI(N)ER}$ in some cases. The differentiation of Seyferts and LI(N)ERs depends on the P_3 dimension. We will further discuss this issue in the following section.

5 DISCUSSIONS

5.1 Presence of LI(N)ERs and their impact

LI(N)ERs are a class of regions that have enhanced low-ionization forbidden lines (e.g. [N II] λ 6583, [S II] $\lambda\lambda$ 6716, 6731, and [O I] λ 6300) compared to Seyferts, but have weaker high-ionization lines (e.g.

[O III]λ5007) (Heckman 1980). Although they are first discovered in the nuclei of galaxies, recent studies show that most galaxies satisfy the line ratio definition for LI(N)ERs have a more extended distribution, and have radial surface brightness profiles and ionization parameter profiles shallower than expected for a compact central source (Sarzi et al. 2006, 2010; Yan & Blanton 2012; Belfiore et al. 2016). The ionization mechanism for LI(N)ERs is still under debate, possible candidates include low-luminosity AGNs, hot evolved stars, shock heating, turbulent mixing layers, and so on (e.g. Slavin, Shull & Begelman 1993; Binette et al. 1994; Dopita & Sutherland 1995; Ho 2008). It is likely that in a small fraction of objects, a small region around the central supermassive black hole, less than tens of parsecs in size, is truly photoionized by the AGN. But the larger scale emission (\sim 100 pc and above), which dominates the line luminosity in most non-SF galaxies, are unlikely to be photoionized by AGNs, given both the energetics (Eracleous, Hwang & Flohic 2010) and spatial profiles (Sarzi et al. 2010; Yan & Blanton 2012).

Current empirical classification of LI(N)ERs is based on emission-line ratios in the BPT diagrams, where clear bimodal distribution of SDSS galaxies in the mixing branch seems to indicate the existence of two different populations (Kewley et al. 2006). Cid Fernandes et al. (2011) introduce another diagnostic based on [N II] λ 6583 and H α emission lines. Their calculation suggests that regions with the equivalent width of H α line smaller than 3 Å are likely to be ionized by hot evolved starts rather than AGNs. More recently, D'Agostino et al. (2019a) use a diagnostic that combines emission line ratios, radial positions, and velocity dispersions to separation shock ionized regions from AGN regions in their IFU sample.

Since LI(N)ERs have relatively low ionization, we expect they are separable from Seyferts in the P_3 direction, as it traces the ionization parameter. The left panel of Fig. 13 shows the distribution of our MaNGA sample on the P_1 – P_3 plane, colour-coded by the equivalent width of the H α line (hereafter EW(H α)). The EW(H α) as a function of the P_3 for spaxels with $P_1 > 0$ is plotted in the right panel. Despite that MaNGA has a small fraction of Seyferts, the transition in EW(H α) along the P_3 dimension is very clear. Viewed on the P_1 – P_2 plane, the Seyferts sit on top of LI(N)ERs and the number density at the AGN locus is dominated by LI(N)ERs. If we exclude spaxels with EW(H α) < 3 Å, as shown in Fig. 14, we still find the remaining spaxels cluster around the best-fitting AGN model, which ensures that the LI(N)ERs do not bias the location of the densest area of the AGN blob. It also indicates that a large portion of the mixing zone could be a product of mixing between LI(N)ERs and SF-ionzed clouds. The SF-ionized clouds will increase the final EW(H α) of the composite spectra, so we expect the 3-Å cut to miss more SF-LI(N)ER composite regions near the SF locus. It would be interesting to compare ionization models for non-AGN LIERs in these diagrams with AGN models in future work.

5.2 SDSS Legacy sample in the new diagram

The separation of LI(N)ERs and Seyferts would be clearer if we use a sample with a larger Seyfert population. Fig. 15 shows the original SDSS-I and II Legacy sample in the P_1 – P_2 plane, as well as the correlation between EW(H α) and positions of galaxies along the P_3 axis for galaxies with $P_1 > 0$. This sample is drawn from the DR7 of SDSS (York et al. 2000; Abazajian et al. 2009). The emission-line measurements were done using an updated version of the code used by Yan et al. (2006), with an additional flux calibration described by Yan (2011). Zero-point corrections to the equivalent width for emission lines were performed according to Yan (2018), in order to remove systematics from inaccurate stellar continuum

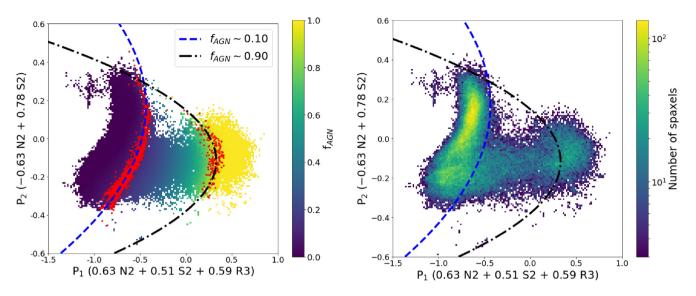


Figure 12. Left-hand panel: map of f_{AGN} in the new diagram. The dashed blue line and the dot-dashed black lines are two fitted demarcation lines that correspond to $f_{AGN} = 0.10$ and 0.90, respectively. The spaxels we select to fit are coloured in red, and we use quadratic functions to fit them. Right-hand panel: density distribution of our sample with the same demarcation lines in the new diagram.

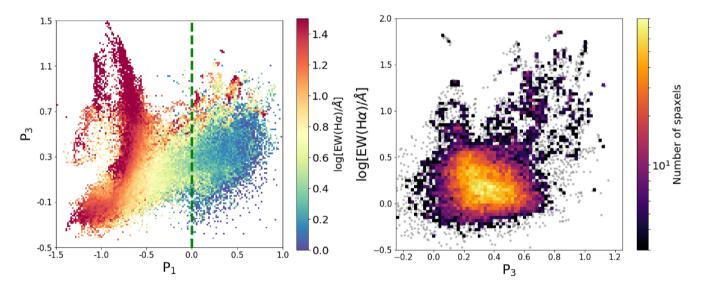


Figure 13. Left-hand panel: map of $log[EW(H\alpha)]$ in the P_1-P_3 projection for MaNGA central spaxels. Right-hand panel: $log[EW(H\alpha)]$ as a function of the positions along the P_3 axis for spaxels with $P_1 > 0.0$, which are to the right of the dashed green line, shown in the left panel.

subtraction. We select only galaxies with the S/N greater than 3 for all emission lines used in this diagram, as we did for the MaNGA sample. The Seyferts and LI(N)ERs now look comparable in number, and there is an obvious separation between them in the P_1 – P_3 plane. The transition from LI(N)ERs to Seyferts also seems slightly lower along the P_3 axis compared with that in the spatially resolved sample of MaNGA.

For galaxies with $P_1 > 0$ in this sample, we can see a continuous increase of EW(H α) as P_3 gets larger. Comparing the right panel of Fig. 15 with that of Fig. 13, we see that the LI(N)ER spaxels in MaNGA with EW(H α) < 3 Å are able to reach higher P_3 values, or equivalently higher ionization parameters, than the SDSS Legacy sample, which is based on single-fibre spectra covering the centre 3 arcsec of each galaxy. It seems to suggest that spatially resolved

LI(N)ERs in MaNGA could have higher ionization. In order to investigate this discrepancy between the two LI(N)ER samples, we select a subsample in MaNGA with $P_1 > 0.0$, EW(H α) < 3 Å and $P_3 > 0.4$, and cross-match it with the SDSS Legacy sample to identify the same galaxies. By selecting MaNGA spaxels within the central 3-arcsec aperture and integrating their emission line flux to mimic the SDSS Legacy observations, we find that the P_3 values derived from the integrated MaNGA data is still higher by 0.16 dex than that derived from the SDSS Legacy spectra for the same set of galaxies. Part of this turns out to be due to the lower S/N in the SDSS Legacy data and our selection criteria of requiring S/N greater than 3 in all emission lines. However, even after we remove the S/N cuts for the emission lines, the median P_3 values of the two samples still differ by 0.11 dex. We think this remaining offset between the

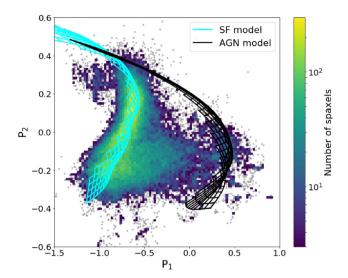


Figure 14. Distribution of MaNGA central spaxels with EW(H α) > 3 Å. Although the composite zone is significantly reduced after the cut, the tip of this zone still traces the AGN model grid.

two sets of data for the same galaxies must be due to a combined effect of the difference in the stellar continuum subtraction, the emission-line measurement algorithm, and the zero-point corrections for the equivalent widths. Using the same data analysis method to process both sample might help to resolve the inconsistency, but this is beyond the scope of this paper and we will leave it to future work.

5.3 Effect of time evolution on SF models

Like most studies that use predictions from photoionization models, we assume the modeled ionized cloud to be in a particular evolutionary stage to match the observations. For example, our SF model is based on a stellar SED that corresponds to a continuous SFH of 4 Myr, and an ionized cloud that has swept the surrounding natal cloud and is exposed to the observer. However, as Pellegrini et al. (2020)

suggested, once the effect of time evolution is considered, a typical H II region can have a wide distribution in the [N II] BPT diagram even at a single metallicity. The variations in line ratios associated with these time-dependent models could be potential sources of systematic uncertainties for our calculations, which treat metallicity as one of the primary indicators of the position of a given H II region in the lineratio space. In the extreme case, zero-age star clusters that are heavily embedded in their natal clouds could produce line ratios similar to those of weak AGNs or DIG due to the internal extinction. However, from the observational point of view, these clouds have very large extinction (with $A_v \gtrsim 3.0$) and cover only a very short period in the evolution history of a typical H II region. Therefore, they are not statistically significant for large samples of H II regions, especially for those with low spatial resolutions, like the MaNGA sample we use (of which the typical PSF ~ 1.5 kpc). The SF locus we observed in this work is more likely to be dominated by clouds that are not heavily impacted by differential extinction, as they usually exhibit much higher optical luminosities.

The general effect of this time evolution could still contribute to the scatter we see in the SF locus. But we note that not all evolutionary stages are detectable for extragalactic observations (see fig. 5 of Pellegrini et al. 2020). Moreover, the data we use in this work are effectively sampling a luminosity-weighted average of different evolutionary stages. Relatively time-steady phases with low internal extinction are more likely to be observed, which is consistent with our model assumptions. Thus, we argue that our interpretation of the SF locus being shaped primarily by the variation in metallicity still holds, at least to the first order. Future work that considers the details of the observational effect on selecting the best-fitting photoionization models would be useful for studying the time-dependent models.

6 CONCLUSIONS

In this paper, we developed a suit of new optical diagnostics different from the original BPT diagnostics, by reprojecting the photoionization models and observational data in a 3D line-ratio space spanned by logarithm of line ratios including N2, S2, and R3.

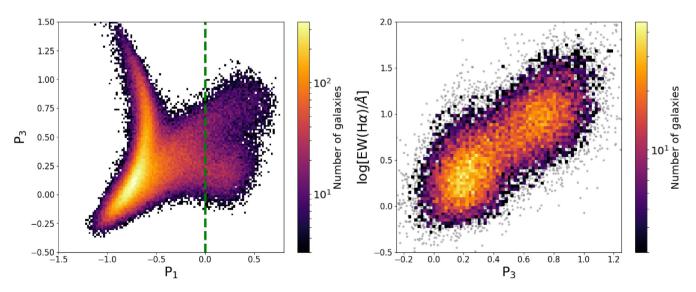


Figure 15. Left-hand panel: density distribution of SDSS galaxies in the P_1 – P_3 projection. Right-hand panel: $\log[EW(H\alpha)]$ as a function of the positions along the P_3 axis for galaxies with $P_1 > 0.0$, which are to the right of the dashed green line shown in the left panel.

Specific angles for the reprojection are obtained according to the intrinsic shapes and orientations of the SF and AGN photoionization model surfaces in the 3D line-ratio space, so that they both appear nearly edge-on and well separated in one of the new projections. The new diagram clearly reveals the spatial relationships between the data and the models, largely removing projection effects inherent to the original BPT diagrams, and guarantees consistency between the data and model in three major line ratios. The applications of the new diagram and the model constraints we obtained can be summarized as the following:

- (1) The new diagram provides clear classification of spectra among the categories of SF, AGN/LI(N)ER, and composites with nearly no ambiguity. Demarcations can be defined with a specified contamination fraction from other sources. Composite objects can be decomposed into SF and AGN/LI(N)ERs with well-constrained fractions.
- (2) With the model surfaces appearing edge-on in our new diagram, the data can put strong constraints on the parameters of photoionization models. We found that the best-fitting SF model requires a continuous star formation history not younger than 4 Myr, and the Dopita13 prescription for N/O abundance ratios, while the best-fitting AGN model needs a power-law energy index of approximately -1.7 between 1 and 100 Ryd, and an N/O prescription significantly higher than that given by Dopita13. This discrepancy in the N/O prescription between the SF and AGN models cannot be resolved by adjusting other parameters.
- (3) With both the SF and AGN models well constrained by the data in our new diagram, it is now possible to derive metallicity for AGN/LI(N)ERs and composite objects. However, the discrepancy in N/O prescription between SF and AGN models prevents us from deriving a consistent metallicity solution. Using MaNGA galaxies within narrow stellar mass bins to define iso-metallicity lines and comparing them with the iso-metallicity lines derived from different model prescriptions, we further demonstrated that the discrepancy in N/O prescriptions cannot be avoided in the current set of models
- (4) LI(N)ERs occupy the same location as Seyferts in our new P_1 – P_2 diagram. But they can be clearly separated from each other in the P_1 – P_3 diagram as P_3 traces ionization parameter approximately. The separation corresponds well with the widely adopted 3-Å cut in EW(H α), although the exact cut in P_1 – P_3 plane seems to depend on the details of the data analysis methods.
- (5) The effect of time evolution of the SF region could potentially contribute to the scatter around the observed SF locus, and introduce systematic uncertainties to our constraints on model parameters. However, the low-resolution extragalactic data we use are likely to be biased towards particular evolutionary stages and thus reduce the scatters.

The lesson we learn from this work is that comparing our models with data in a higher dimensional space composed by multiple line ratios can reveal a great deal of information and put strong constraints on the models. Even though higher dimensional space is difficult to visualize, certain projections of it could provide valuable intuitions. Those intuitions could facilitate the construction of algorithms to quantitatively evaluate our models.

ACKNOWLEDGEMENTS

The authors wish to thank David Law for useful comments on a draft of this paper. We acknowledge support by NSF AST-1715898 and NASA grant 80NSSC20K0436 subaward S000353.

Funding for the Sloan Digital Sky Survey IV has been provided by the Alfred P. Sloan Foundation, the US Department of Energy Office of Science, and the Participating Institutions. SDSS acknowledges support and resources from the Center for High-Performance Computing at the University of Utah. The SDSS website is www.sdss.org.

SDSS is managed by the Astrophysical Research Consortium for the Participating Institutions of the SDSS Collaboration including the Brazilian Participation Group, the Carnegie Institution for Science, Carnegie Mellon University, the Chilean Participation Group, the French Participation Group, Harvard-Smithsonian Center for Astrophysics, Instituto de Astrofísica de Canarias, The Johns Hopkins University, Kavli Institute for the Physics and Mathematics of the Universe (IPMU)/University of Tokyo, the Korean Participation Group, Lawrence Berkeley National Laboratory, Leibniz Institut für Astrophysik Potsdam (AIP), Max-Planck-Institut für Astronomie (MPIA Heidelberg), Max-Planck-Institut für Astrophysik (MPA Garching), Max-Planck-Institut für Extraterrestrische Physik (MPE), National Astronomical Observatories of China, New Mexico State University, New York University, University of Notre Dame, Observatório Nacional/MCTI, The Ohio State University, Pennsylvania State University, Shanghai Astronomical Observatory, United Kingdom Participation Group, Universidad Nacional Autónoma de México, University of Arizona, University of Colorado Boulder, University of Oxford, University of Portsmouth, University of Utah, University of Virginia, University of Washington, University of Wisconsin, Vanderbilt University, and Yale University.

DATA AVAILABILITY

The observational data underlying this paper are publicly available in the DR15¹ and DR7² of SDSS (see Sections 2 and 5.2). The derived data generated in this research and data concerning the photoionization models will be shared on reasonable request to the corresponding author.

REFERENCES

Abazajian K. N. et al., 2009, ApJS, 182, 543

Baldwin J. A., Phillips M. M., Terlevich R., 1981, PASP, 93, 5

Belfiore F. et al., 2016, MNRAS, 461, 3111

Belfiore F. et al., 2017, MNRAS, 469, 151

Belfiore F. et al., 2019, AJ, 158, 160

Binette L., Magris C. G., Stasińska G., Bruzual A. G., 1994, A&A, 292, 13

Blanc G. A., Kewley L., Vogt F. P. A., Dopita M. A., 2015, ApJ, 798, 99

Blanton M. R. et al., 2017, AJ, 154, 28

Bundy K. et al., 2015, ApJ, 798, 7

Cappellari M., 2017, MNRAS, 466, 798

Cappellari M., Emsellem E., 2004, PASP, 116, 138

Cid Fernandes R., Stasińska G., Mateus A., Vale Asari N., 2011, MNRAS, 413, 1687

Cowie L. L., Songaila A., 1986, ARA&A, 24, 499

D'Agostino J. J., Kewley L. J., Groves B. A., Medling A., Dopita M. A., Thomas A. D., 2019a, MNRAS, 485, L38

D'Agostino J. J., Kewley L. J., Groves B., Byler N., Sutherland R. S., Nicholls D., Leitherer C., Stanway E. R., 2019b, ApJ, 878, 2

De Robertis M. M., Dufour R. J., Hunt R. W., 1987, JRASC, 81, 195

Dopita M. A., Sutherland R. S., 1995, ApJ, 455, 468

Dopita M. A., Kewley L. J., Heisler C. A., Sutherland R. S., 2000, ApJ, 542, 224

¹www.sdss.org/dr15/data_access/.

²classic.sdss.org/dr7/access/.

Dopita M. A., Groves B. A., Sutherland R. S., Binette L., Cecil G., 2002, ApJ, 572, 753

Dopita M. A., Sutherland R. S., Nicholls D. C., Kewley L. J., Vogt F. P. A., 2013, ApJS, 208, 10

Dopita M. A., Kewley L. J., Sutherland R. S., Nicholls D. C., 2016, Ap&SS, 361, 61

Drory N. et al., 2015, AJ, 149, 77

Eracleous M., Hwang J. A., Flohic H. M. L. G., 2010, ApJ, 711, 796

Ferland G. J. et al., 2017, Rev. Mex. Astron. Astrofis., 53, 385

Goddard D. et al., 2017, MNRAS, 466, 4731

Grevesse N., Asplund M., Sauval A. J., Scott P., 2010, Ap&SS, 328, 179

Groves B. A., Dopita M. A., Sutherland R. S., 2004a, ApJS, 153, 9

Groves B. A., Dopita M. A., Sutherland R. S., 2004b, ApJS, 153, 75

Gunn J. E. et al., 2006, AJ, 131, 2332

Heckman T. M., 1980, A&A, 500, 187

Ho L. C., 2008, ARA&A, 46, 475

Ho L. C., Filippenko A. V., Sargent W. L. W., 1997, ApJS, 112, 315

Jenkins E. B., 1987, in Hollenbach D. J., Thronson H. A., Jr, eds, Astrophysics and Space Science Library, Vol. 134, Interstellar Processes. Springer, Berlin, p. 533

Ji X., Yan R., Riffel R., Drory N., Zhang K., 2020, MNRAS, 496, 1262 Kauffmann G. et al., 2003, MNRAS, 346, 1055

Kennicutt Robert C. J., Bresolin F., Garnett D. R., 2003, ApJ, 591, 801

Kewley L. J., Dopita M. A., 2002, ApJS, 142, 35

Kewley L. J., Dopita M. A., Sutherland R. S., Heisler C. A., Trevena J., 2001, ApJ, 556, 121

Kewley L. J., Groves B., Kauffmann G., Heckman T., 2006, MNRAS, 372, 961

Law D. R. et al., 2015, AJ, 150, 19

Law D. R. et al., 2016, AJ, 152, 83

Leitherer C. et al., 1999, ApJS, 123, 3

Lusso E. et al., 2010, A&A, 512, A34

Mouhcine M., Contini T., 2002, A&A, 389, 106

Nicholls D. C., Sutherland R. S., Dopita M. A., Kewley L. J., Groves B. A., 2017, MNRAS, 466, 4403

Pellegrini E. W., Rahner D., Reissl S., Glover S. C. O., Klessen R. S., Rousseau-Nepton L., Herrera-Camus R., 2020, MNRAS, 496, 339

Sarzi M. et al., 2006, MNRAS, 366, 1151

Sarzi M. et al., 2010, MNRAS, 402, 2187

Schaefer A. L., Tremonti C., Belfiore F., Pace Z., Bershady M. A., Andrews B. H., Drory N., 2020, ApJ, 890, L3

Shaw R. A., Dufour R. J., 1995, PASP, 107, 896

Slavin J. D., Shull J. M., Begelman M. C., 1993, ApJ, 407, 83

Smee S. A. et al., 2013, AJ, 146, 32

Storchi-Bergmann T., Schmitt H. R., Calzetti D., Kinney A. L., 1998, AJ, 115, 909

Thomas A. D., Kewley L. J., Dopita M. A., Groves B. A., Hopkins A. M., Sutherland R. S., 2019, ApJ, 874, 100

Tremonti C. A. et al., 2004, ApJ, 613, 898

van Zee L., Salzer J. J., Haynes M. P., O'Donoghue A. A., Balonek T. J., 1998, AJ, 116, 2805

Veilleux S., Osterbrock D. E., 1987, ApJS, 63, 295

Vila Costas M. B., Edmunds M. G., 1993, MNRAS, 265, 199

Vogt F. P. A., Dopita M. A., Kewley L. J., Sutherland R. S., Scharwächter J., Basurah H. M., Ali A., Amer M. A., 2014, ApJ, 793, 127

Wake D. A. et al., 2017, AJ, 154, 86

Westfall K. B. et al., 2019, AJ, 158, 231

Yan R. et al., 2016a, AJ, 151, 8

Yan R. et al., 2016b, AJ, 152, 197

Yan R., 2011, AJ, 142, 153

Yan R., 2018, MNRAS, 481, 467

Yan R., Blanton M. R., 2012, ApJ, 747, 61

Yan R., Newman J. A., Faber S. M., Konidaris N., Koo D., Davis M., 2006, ApJ, 648, 281

York D. G. et al., 2000, AJ, 120, 1579

SUPPORTING INFORMATION

Supplementary data are available at *MNRAS* online.

fig1_leftp.mp4

Please note: Oxford University Press is not responsible for the content or functionality of any supporting materials supplied by the authors. Any queries (other than missing material) should be directed to the corresponding author for the article.

This paper has been typeset from a TEX/LATEX file prepared by the author.