The SAMI galaxy survey: gas velocity dispersions in low-*z* star-forming galaxies and the drivers of turbulence

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ABSTRACT

We infer the intrinsic ionized gas kinematics for 383 star-forming galaxies across a range of integrated star formation rates (SFR $\in [10^{-3}, 10^2] M_{\odot} \text{ yr}^{-1}$) at $z \leq 0.1$ using a consistent 3D forward-modelling technique. The total sample is a combination of galaxies from the Sydney-AAO Multiobject Integral field Spectrograph (SAMI) Galaxy survey and DYnamics of Newly Assembled Massive Objects survey. For typical low-*z* galaxies taken from the SAMI Galaxy Survey, we find the vertical velocity dispersion ($\sigma_{v,z}$) to be positively correlated with measures of SFR, stellar mass, H I gas mass, and rotational velocity. The greatest correlation is with SFR surface density (Σ_{SFR}). Using the total sample, we find $\sigma_{v,z}$ increases slowly as a function of integrated SFR in the range SFR $\in [10^{-3}, 1] M_{\odot} \text{ yr}^{-1}$ from 17 ± 3 to $24 \pm 5 \text{ km s}^{-1}$ followed by a steeper increase up to $\sigma_{v,z} \sim 80 \text{ km s}^{-1}$ for SFR $\gtrsim 1 M_{\odot} \text{ yr}^{-1}$. This is consistent with recent theoretical models that suggest a $\sigma_{v,z}$ floor driven by star formation feedback processes with an upturn in $\sigma_{v,z}$ at higher SFR driven by gravitational transport of gas through the disc.

Key words: methods: data analysis – methods: statistical – techniques: imaging spectroscopy – galaxies: disc.

1 INTRODUCTION

Galaxies at z > 1 typically have velocity dispersions greater than nearby galaxies (Kassin et al. 2012; Wisnioski et al. 2015; Johnson

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(Epinat, Amram & Marcelin 2008; Moiseev, Tikhonov & Klypin 2015; Varidel et al. 2016; Yu et al. 2019). Although this has been observed, the process by which galaxies settle to lower velocity dispersions across epochs is not well understood.

Another important observation is that galaxies at all epochs exhibit velocity dispersions that are greater than expected by the thermal contribution of the gas alone. In the case of ionized gas measured using the H α emission line, the characteristic temperature of 10⁴ K corresponds to an expected velocity dispersion of ~9 km s⁻¹ (Glazebrook 2013). Galaxies have velocity dispersions >9 km s⁻¹ at all epochs.

Studies suggest that turbulent motions above the thermal contribution dissipate on time-scales of the order of the flow crossing time (Mac Low et al. 1998; Stone, Ostriker & Gammie 1998; Mac Low 1999). The crossing time for a galaxy with Toomre stability (Toomre 1964) of $Q \sim 1$ will be of order the dynamical time, which is typically $\mathcal{O}(100 \text{ Myr})$; Krumholz et al. 2018). If the turbulent motions are on the scale of giant molecular clouds (GMCs), it will decay on $\mathcal{O}(<10 \text{ Myr})$. Therefore, we should rarely see galaxies with velocity dispersions greater than the thermal contribution, unless there is an ongoing driving mechanism to sustain the observed gas turbulence.

Numerous energetic sources have been proposed to contribute to the non-thermal turbulence observed in galaxies. These drivers can typically be split into star formation feedback processes (Norman & Ferrara 1996; Mac Low & Klessen 2004; Krumholz & Matzner 2009; Murray, Quataert & Thompson 2010), gravitational transport of gas on to (Elmegreen & Burkert 2010; Hopkins, Kereš & Murray 2013) or through (Krumholz & Burkert 2010) the disc, dynamical drivers such as shear and differential rotations across the disc (Federrath et al. 2016, 2017), or interactions between galaxy components (e.g. Dobbs & Bonnell 2007; Dekel et al. 2009; Aumer et al. 2010; Ceverino, Dekel & Bournaud 2010; Oliva-Altamirano et al. 2018). In this paper, we will be focusing primarily on differentiating star formation feedback processes and gravitational transport of gas through the disc due to the clear predictions that have been made in the integrated star formation rate (SFR) and global velocity dispersion (σ_v) plane (Krumholz & Burkhart 2016; Krumholz et al. 2018).

Star formation feedback is thought to be dominated by the energy imparted by supernovae (Norman & Ferrara 1996; Mac Low & Klessen 2004). However, other drivers such as stellar winds, expansion of H II regions (Chu & Kennicutt 1994; Matzner 2002), and radiation pressure in high-density star clusters (Krumholz & Matzner 2009; Murray et al. 2010) will also inject momentum into the interstellar medium (ISM). Observational evidence for star formation feedback as the primary driver of gas turbulence has been argued by observing that SFR is correlated with σ_v . The SFR– σ_v correlation has been shown both within a single sample at constant redshift (Green et al. 2010, 2014; Moiseev et al. 2015; Yu et al. 2019) and by combining multiple samples across epochs (Green et al. 2010, 2014).

Assuming that star formation feedback processes are a significant driver of the turbulence, it would be natural to expect a relation between local SFR surface density (Σ_{SFR}) and local velocity dispersion. There are conflicting results in the literature regarding the relationship between these local quantities. Some studies have found a significant relationship (Lehnert et al. 2009, 2013), whereas others have found the localized relationship to be weak (Genzel et al. 2011; Varidel et al. 2016; Zhou et al. 2017; Übler et al. 2019).

Furthermore, the physical mechanism for an energetic source to account for velocity dispersions due to star formation feedback of several 10s of km s⁻¹ is not well established. Constructing equilibrium solutions between gravitational infall of the disc supported by outward pressure solely by supernovae leads to $\sigma_v \leq 25$ km s⁻¹ with little variation as a function of SFR (Ostriker & Shetty 2011; Krumholz et al. 2018). An alternative approach that can account for increased turbulence is to assume that the star formation efficiency per free-fall time ($\epsilon_{\rm ff}$) changes as a function of galaxy properties, thus changing the energetic input from star formation feedback processes (Faucher-Giguère, Quataert & Hopkins 2013). However, numerous observations suggest that $\epsilon_{\rm ff}$ is approximately constant across a wide range of galaxy properties (Krumholz & Tan 2007; Krumholz, Dekel & McKee 2012; Federrath 2013; Salim, Federrath & Kewley 2015; Krumholz, McKee & Bland -Hawthorn 2019).

An alternative set of driving mechanisms is due to gravitational effects. This includes the initial gravitationally unstable formation of the disc (Aumer et al. 2010), which can account for short-lived supersonic turbulence on the order of the disc formation time, $\mathcal{O}(100 \text{ Myr})$. It is thought that the supersonic turbulence that is initially set at disc formation can be maintained by the gravitational transport of gas through the disc (Krumholz & Burkert 2010). Krumholz & Burkhart (2016) also argued that the gravitational transport model predicts an increase in velocity dispersion at increased SFR that is more consistent with the data than models assuming star formation feedback processes.

A further complication involved in inferring the ongoing drivers of turbulence across epochs is the effects of the spectral and spatial resolution on the observed velocity dispersion. The spectral resolution broadens the observed emission line often on order of the intrinsic velocity dispersion. This is typically accounted for by convolving the modelled emission line profile by the known line-spread function (LSF) while fitting to the data (e.g. Förster Schreiber et al. 2009; Davies et al. 2011; Green et al. 2014; Varidel et al. 2019). This is a reasonable approximation as long as the model assumptions regarding the LSF are well known.

The spatial resolution is more difficult to account for as it acts to blur the emission line flux spatially per spectral slice. The observed velocity dispersion is then a complex function of the intrinsic flux distribution, line-of-sight (LoS) velocity profile, and LoS velocity dispersion profile. This effect is usually referred to as beam smearing.

In general, beam smearing acts to increase the observed velocity dispersion particularly where the velocity gradient is steepest (Davies et al. 2011; Glazebrook 2013), and in detail can result in spurious substructure in the velocity dispersion profile (Varidel et al. 2019). Furthermore, beam smearing could result in spurious correlations such as the SFR– σ_v correlation, as SFR is related to the mass thatpes the gravitational potential, and thus increases the velocity gradient at the centre of galaxies with higher SFR. Similarly, the width of the point spread function (PSF) relative to the galaxy size increases for increasing *z*, thus resulting in higher observed velocity dispersions if beam smearing is not corrected for appropriately.

The SFR– σ_v relation has been used to distinguish between the different energetic sources of turbulence (Krumholz & Burkhart 2016; Krumholz et al. 2018). However, comparisons between theoretical models and observations have typically been performed by combining several studies with different redshift ranges and beam smearing corrections. In this paper, we improve comparisons of the observed velocity dispersion to theoretical models by studying a sample of nearby galaxies using a single technique to mitigate the effects of beam smearing. The data encompass a wide range of SFR $\in [10^{-3}, 10^2] \text{ M}_{\odot} \text{ yr}^{-1}$ of local galaxies at $z \leq 0.1$. The

combined sample is comprised of observations from the Sydney-AAO Multiobject Integral field Spectrograph (SAMI) Galaxy Survey Data Release Two (SAMI Galaxy Survey DR2; Croom et al. 2012; Scott et al. 2018) and the DYnamics of Newly Assembled Massive Objects (DYNAMO; Green et al. 2014) survey. We use a consistent disc-fitting routine referred to as BLOBBY3D (Varidel et al. 2019), for all the galaxy gas kinematic modelling in this paper. BLOBBY3D is a disc fitting code that constructs a regularly rotating thin-disc galaxy model in 3D (position–position–wavelength space) that is then convolved by the PSF and LSF prior to comparing the model to the data. In that way, it can account for the effect of beam smearing when inferring the velocity dispersion of the galaxy.

The outline of this paper is as follows. In Section 2, we describe the SAMI Galaxy Survey and DYNAMO surveys, as well as our sample selection criteria. In Section 3, we outline the methods used to measure the key gas kinematic properties. In Section 4, we will discuss our results. In Section 5, we compare our results to theoretical models of the drivers for turbulence. We summarize our conclusions in Section 6. Throughout this paper, we assume the concordance cosmology ($\Omega_{\Lambda} = 0.7$, $\Omega_m = 0.3$, $H_0 = 70 \,\mathrm{km \, s^{-1} \, Mpc^{-1}}$; Hinshaw et al. 2009) and a Chabrier (2003) initial mass function (IMF).

2 DATA SELECTION

2.1 The SAMI Galaxy Survey

The SAMI Galaxy Survey was conducted with the SAMI (Croom et al. 2012). SAMI was mounted at the Anglo-Australian Telescope (AAT) that provided a 1° diameter field of view (FoV). SAMI used 13 fused fibre bundles, known as Hexabundles (Bland-Hawthorn et al. 2011; Bryant et al. 2014), with a 75 per cent fill factor. Each bundle contains 61 fibres of 1.6 arcsec diameter, resulting in an approximately 15 arcsec diameter FoV. The integral-field units (IFUs) as well as 26 sky fibres were attached to pre-drilled plates using magnetic connectors. SAMI fibres were fed to the double-beam AAOmega spectrograph (Sharp et al. 2006). The 580V grating at 3750–5750 Å provides a resolution of R = 1808 ($\sigma = 70.4$ km s⁻¹ at 4800 Å) and the 1000R grating from 6300–7400 Å providing a resolution of R = 4304 ($\sigma = 29.6$ km s⁻¹ at 6850 Å; Scott et al. 2018).

During the survey, observations of over 3000 galaxies were obtained. Target selection for the SAMI Galaxy Survey is provided in Bryant et al. (2015). The redshift range for the observed galaxies was 0.004 < z < 0.113 and a stellar mass range of $7.5 < \log (M_*/M_{\odot}) < 11.6$. The full width at half-maximum (FWHM) of the seeing distribution was 1.10 arcsec < FWHM_{PSF} < 3.27 arcsec. Relevant data used for the analysis in this paper are from the SAMI Galaxy Survey DR2 (Scott et al. 2018). This includes the aperture spectra, emission line products (Green et al. 2018), data cubes (Sharp et al. 2015), and input catalogue (Bryant et al. 2015).

2.2 Sample selection from the SAMI Galaxy Survey

Our aim was to select galaxies on the star-forming main sequence within the SAMI Galaxy Survey. As such, we performed the following selection criteria cuts to the sample from the SAMI Galaxy Survey DR2 (Scott et al. 2018).

Star-forming galaxies are selected by applying a cut-off integrated H α equivalent width of EW > 3 Å (Cid Fernandes et al. 2011). The equivalent width is calculated as the total H α flux compared to the total continuum flux across the SAMI FoV. The continuum flux in the region around H α is estimated by calculating the mean continuum in the wavelength range [6500, 6540] Å. The integrated H α flux estimates is sourced from the SAMI Galaxy Survey DR2 emission line data products.

We remove galaxies with ionized emission from non-star-forming sources such as active galactic nuclei (AGN) and low-ionization nuclear emission-line regions (LINERs). To implement these criteria, we remove galaxies where the AGN classification criteria proposed by Kauffmann et al. (2003) is met:

$$\log([O \text{ III}]/\text{H}\beta) > \frac{0.61}{\log([N \text{ II}]/\text{H}\alpha) - 0.05} + 1.3.$$
(1)

[O III] and [N II] represent the emission line fluxes at 5007 and 6583 Å, respectively. The line fluxes are estimated for the central region of the galaxy where AGN and LINER contamination should be greatest, using the 1.4 arcsec aperture spectra from the SAMI Galaxy Survey DR2.

We retain galaxies that are face-on up to e = 1 - b/a = 0.5 (0° $< i < 60^{\circ}$, assuming a thin disc). We avoid galaxies observed at high inclination as the intrinsic velocity dispersion is more difficult to constrain due to beam smearing. Plus galaxies are optically thick such that edge-on observations limit the ability to observe the integrated LoS from the entire galaxy. Furthermore, a thin disc model is assumed in BLOBBY3D, such that the galaxies will not be well modelled when observed close to edge-on.

We apply the following signal-to-noise cut on the spaxels in the data. We first apply a mask to spaxels with H α flux signal to noise <3. Spatially resolved H α flux and its error are obtained from the SAMI Galaxy Survey DR2 pipeline. We then construct groups of unmasked spaxels that are adjacent and meet the signalto-noise criteria. The largest unmasked group is retained, whereas the remaining spaxels are masked. We retain galaxies that had at least 300 unmasked spaxels.

The above masking routine only finds the largest group of spaxels, which in principle could reject clumpy flux profiles. In practice, the effect of removing H α clumps originating from the galaxy was negligible. Instead, it primarily removed spurious spaxels that were reported to have high signal to noise, yet by eye did not appear to be legitimate detections of flux originating from the galaxy.

We also remove mergers or galaxies with clearly disturbed gas kinematics from our final sample. Potential mergers were determined by eye from observations of the gas kinematic maps. In total, 9 galaxies were removed from our final sample due to these criteria.

There are 1523 galaxies in the SAMI Galaxy Survey DR2 where all of the above diagnostic criteria are measurable. 342 galaxies remain once our criteria is applied. Fig. 1 shows that we are selecting galaxies along the star-forming main sequence. We see a clear bimodal distribution in the log equivalent width, where we have selected those galaxies with EW > 3 Å. The equivalent width cut removes massive galaxies that are typically passive, which can be seen when plotting the equivalent width compared to M_{*} and R_e. There are a limited number of galaxies in our sample with 3 Å < $EW \leq 10$ Å as many of those galaxies are removed due to being classified as LINER/AGN or having < 300 spaxels that meet our signal-to-noise masking criteria.

Removing highly inclined galaxies results in a large cut to our sample, but does not bias our sample along any galaxy properties. Also, the selection of galaxies with at least 300 unmasked spaxels does remove galaxies with $R_e \leq 1$ arcsec, but there are very few of these galaxies in the underlying SAMI Galaxy Survey DR2 sample.



Figure 1. Galaxy parameters for our sample of 342 galaxies (red) selected from the total SAMI Galaxy Survey (grey). We show the marginalized (diagonal) and conditional (off-diagonal) distributions for the stellar mass ($\log_{10}(M^*/M_{\odot})$), effective radius ($\log_{10}(R_e/asec)$), ellipticity (e = 1 - b/a), H α equivalent width ($\log_{10}(EW/Å)$), and NSNGT3. NSNGT3 corresponds to the number of spaxels that meet our signal-to-noise masking criteria. We select a sample of star-forming galaxies from the SAMI Galaxy Survey with inclination and signal-to-noise cuts that can be adequately modelled using BLOBBY3D.

2.3 DYNAMO sample

The DYNAMO (Green et al. 2014) survey consists of a sample of star-forming galaxies in the local Universe ($z \leq 0.1$). These galaxies were classified as star-forming in the MPA-JHU Value Added Catalog from the Sloan Digital Sky Survey (SDSS; York et al. 2000). The galaxies comprising the DYNAMO survey were chosen primarily based on H α luminosity. The aim was to include both high H α luminous galaxies, which are rare in the local Universe, as well as a sample of typical galaxies in the local Universe. The resulting galaxy sample ranged SFR $\in [1, 100] M_{\odot} \text{ yr}^{-1}$.

The data for the DYNAMO samples was obtained via observations using the 3.9 m AAT and the ANU 2.3 m Telescope

at Siding Spring Observatory. The AAT was equipped with the SPIRAL IFU with the AAOmega Spectrograph (Sharp et al. 2006). SPIRAL is an array of 32×16 square, 0.7 arcsec lenslets with a contiguous integral field of 22.4 arcsec $\times 11.2$ arcsec. The 1700I grating was used on the red spectrograph providing a nominal resolution power of $R \sim 12000$. The ANU 2.3 m Telescope was equipped with the Wide-Field Spectrograph (WiFeS; Dopita et al. 2007). WiFeS has a 25 arcsec \times 38 arcsec FoV with either 1.0 arcsec \times 0.5 arcsec or 1.0 arcsec \times 1.0 arcsec spaxels. The I7000 grating was chosen for the red arm, which has a 6893–9120 Å wavelength range with a spectral resolving power of $R \sim$ 7000.

A total of 67 galaxies comprised the original DYNAMO sample. We remove galaxies observed at $i > 60^\circ$, where *i* has been measured using the SDSS photometric pipeline using an exponential disc fit to the *r* band. We perform the same masking criteria as described for the galaxies from the SAMI Galaxy Survey. We also remove galaxies with less than 30 unmasked spaxels. A total of 41 galaxies were retained from the original DYNAMO sample.

3 METHODS

3.1 Modelling the gas disc kinematics

We use BLOBBY3D (Varidel et al. 2019) to infer the intrinsic gas kinematics for the observed galaxies. BLOBBY3D is a forward-fitting disc modelling procedure. It assumes that the gas lies in a regularly rotating thin disc. The prior for the spatial gas distribution within the disc allows for clumpy gas profiles using a hierarchical Gaussian mixture model. The model is constructed in 3D (position–position–wavelength space) and then convolved in accordance with the PSF and instrumental broadening by the LSF. The convolved model is then compared to the observed data cube.

The advantage of BLOBBY3D is that it is capable of performing inference for the spatial gas distribution, including substructure, plus the gas kinematics simultaneously. This is important as the effect of beam smearing is a function of the spatial gas distribution being blurred per spectral slice. As such, the observed gas kinematics is a complex function of the intrinsic spatial gas distribution, the velocity profile, and the velocity dispersion plus instrumental broadening and beam smearing. For example, Varidel et al. (2019) found that it is possible to observe spurious substructure in the gas kinematics in a symmetric regularly rotating disc with an asymmetric spatial gas distribution plus beam smearing.

Previous testing of BLOBBY3D has found that it is well optimized to infer the intrinsic velocity dispersion of galaxies (Varidel et al. 2019). BLOBBY3D was compared to an alternative forward-fitting methodology known as ^{3D}BAROLO (Di Teodoro & Fraternali 2015). It was also compared to other heuristic modelling approaches that have been used in the literature, such as estimating the velocity dispersion in the outskirts of the galaxy (e.g. Zhou et al. 2017), correcting the observed velocity dispersion as a function of the velocity gradient (e.g. Varidel et al. 2016), and subtracting the velocity gradient in quadrature from the observed velocity dispersion (e.g. Oliva-Altamirano et al. 2018). BLOBBY3D was found to infer the intrinsic velocity dispersion more accurately than these alternative methods, particular for galaxies where the PSF or velocity gradient were greatest.

The parametrization for BLOBBY3D is set within the Bayesian framework. The joint prior distribution for the parameters, hyperparameters, and data were defined in Varidel et al. (2019). We only make minor changes to the priors that were previously proposed. We outline the motivation for changing some of the prior distributions next.

The joint prior distribution used for this work performs inferences for the H α flux plus the [N II]/H α emission flux ratio for each spatial Gaussian flux profile (often referred to as a 'blob' in BLOBBY3D). The gas kinematics have been assumed to be consistent across the different gas components. Therefore, the inferences for the kinematics are constrained using extra information from the [N II] emission lines at 6548.1 and 6583.1 Å. The ratio of the flux between the [N II] emission lines is assumed to be $F_{6583.1}/F_{6548.1} = 3$.

To simplify the inference for the velocity dispersion, we assume a constant velocity dispersion across the disc $(\sigma_{v,0})$. We assume

no radial gradient as the results for some galaxies returned large positive gradients when using the prior suggested by Varidel et al. (2019). The large spatial gradients in velocity dispersion after convolution appeared to be overfitting for wider tailed non-Gaussian emission line profiles. Therefore, we removed the velocity dispersion gradient from the inference in order to robustly infer the constant velocity dispersion component for the large sample of galaxies that were studied in this work.

We have also widened the bounds for our priors for the systemic velocity (v_{sys}) and the asymptotic velocity (v_c) in order to model a larger set of galaxies than was performed by Varidel et al. (2019). Our new priors are

 $v_{\rm sys} \sim {\rm Cauchy}(0, 30 \,{\rm km \, s^{-1}})T(-300 \,{\rm km \, s^{-1}}, 300 \,{\rm km \, s^{-1}}),$ (2)

 $v_c \sim \text{Loguniform}(1 \text{ km s}^{-1}, 1000 \text{ km s}^{-1}),$

where T(a, b) represents the distribution being truncated to the interval [a, b].

3.1.1 Mitigating the effects of beam smearing

The effect of beam smearing by the PSF is accounted for in BLOBBY3D by convolving the underlying model constructed by the PSF, prior to calculating the likelihood function. The PSF profile assumed in BLOBBY3D is a superposition of 2D concentric circular Gaussian profiles. Therefore, the PSF needs to first be modelled assuming this flux profile.

The SAMI Galaxy Survey pipeline provides estimates for the PSF by fitting a profile to a star that was observed simultaneously with the galaxy. We have used the Moffat profile estimates, where the PSF is described as

$$p(r) = \frac{\beta - 1}{\pi \alpha^2} \left(1 + \frac{r^2}{\alpha^2} \right)^{-\beta}.$$
(4)

 α is the FWHM and β is a shape parameter that controls the tails of the Moffat profile.

To refactor the Moffat profile parameters into a set of concentric Gaussians, we construct the 1D Moffat profile, then fit it with two 1D Gaussians. Two Gaussians were enough to adequately model the PSF profile. The estimated Gaussian parameters are then passed to BLOBBY3D.

For the DYNAMO sample, the FWHM of the PSF was measured during observations. As such, we assumed a 2D circular Gaussian profile to be representative of the PSF for the DYNAMO sample. Thus, the underlying model in BLOBBY3D was convolved with a Gaussian profile prior to comparing the model to the data for our galaxies from the DYNAMO survey.

3.1.2 Continuum substraction

BLOBBY3D requires the data to be continuum subtracted. For galaxies from the SAMI Galaxy Survey, we use the continuum models made available in the SAMI Galaxy Survey DR2 pipeline. The full description for the continuum modelling routine is described in Owers et al. (2019). We estimate the continuum for the galaxies from the DYNAMO survey using a 300 bin moving median filter as also implemented by Green et al. (2014).

It is possible for the continuum modelling to introduce systematics in the resulting continuum subtracted data cube. These systematics may not be well accounted for in the BLOBBY3D approach. We make the assumption that the stellar continuum will

(3)



Figure 2. Comparing the PSF Moffat profile parameters α and β to the inferred global velocity dispersion for galaxies in our sample from the SAMI Galaxy Survey. We also show the PDF of the Spearman-rank correlation coefficients estimated using 10⁴ bootstrap samples (bottom). $\rho = 0$ lies within the 68 per cent shortest credible intervals suggesting that $\sigma_{v,0}$ is adequately corrected for beam smearing.

be adequately modelled in regions of high H α signal to noise. This is a significant motivation for implementing the H α signal-to-noise masking outlined in Section 2.2.

3.1.3 Posterior optimization

We use DNEST4 (Brewer, Pártay & Csányi 2011; Brewer & Foreman-Mackey 2018) to get a point estimate of the maxima for the posterior Probability Density Function (PDF). DNEST4 is a sampling algorithm based on nested sampling (Skilling 2004), where the new levels are constructed by exploring a weighted mixture of the previous levels. Exploration of the levels is performed using a Metropolis Markov Chain Monte Carlo. The multilevel exploration allows DNEST4 to be significantly more robust to local maxima compared to typical nested sampling, allowing for the exploration of high parameter spaces and multimodal posterior distributions. Estimated values throughout this paper are of the maximum posterior PDF value in the chain sampled using DNEST4.

3.2 Global velocity dispersion

3.2.1 Beam smearing corrections

Assuming that BLOBBY3D accurately corrects for beam smearing, there should be no residual correlation between the PSF profile parameters and the inferred intrinsic velocity dispersion ($\sigma_{v,0}$). The distribution of $\sigma_{v,0}$ is consistent with our expectations for a beam smearing corrected sample. Fig. 2 shows a comparison between the PSF Moffat profile parameters and $\sigma_{v,0}$ for our sample from the SAMI Galaxy Survey. For both α and β , zero remains inside the 68 per cent shortest credible intervals for the Spearman-rank correlation coefficients.



Figure 3. $\sigma_{v,0}$ estimated using BLOBBY3D compared to the arithmetic mean of the single-component fits per spaxel ($\sigma_{v,\text{uncorrected}}$) to each galaxy from the SAMI Galaxy Survey sample. Estimates for the velocity dispersion are typically lower using BLOBBY3D as it mitigates the effects of beam smearing.

For galaxies from the DYNAMO survey, the Spearman-rank correlation coefficient is estimated as $\rho(\text{FWHM}, \sigma_v) = 0.10^{+0.17}_{-0.17}$. As zero remains within the 68 per cent confidence interval, this result is also consistent with a beam smearing corrected sample.

We also compare $\sigma_{v,0}$ to an estimate of the velocity dispersion that was not corrected for beam smearing ($\sigma_{v, \text{uncorrected}}$). The uncorrected estimator is calculated as the arithmetic mean velocity dispersion across the FoV, when fitting a single Gaussian component to each spaxel. Spaxels with H α signal to noise < 3 are masked in this process to eliminate the effects of poorly constrained spaxels on the final estimate.

Estimates for $\sigma_{v,0}$ are significantly lower than $\sigma_{v,\text{uncorrected}}$ (see Fig. 3). Using the sample of galaxies from the SAMI Galaxy Survey, typical corrections were $\Delta \sigma_v = -5.3^{+4.0}_{-7.0} \text{ km s}^{-1}$ and $\Delta \sigma_v / \sigma_{v,0} = -0.20^{+0.14}_{-0.18}$, where $\Delta \sigma_v = \sigma_{v,0} - \sigma_{v,\text{uncorrected}}$. The typical beam smearing corrections are consistent with the results found by Varidel et al. (2019) on a sample of 20 star-forming galaxies in the SAMI Galaxy Survey using BLOBBY3D.

All estimated values have $\sigma_{v,0} > \sigma_{v,\text{thermal}} = 9 \text{ km s}^{-1}$. $\sigma_{v,\text{thermal}}$ is the typical emission line width expected for a H II region at $\sim 10^4$ K (Glazebrook 2013). As such, $\sigma_{v,\text{thermal}}$ sets a physically motivated lower bound.

3.2.2 Considerations of the effects of the LSF on the velocity dispersion estimates

The SAMI instrument has the spectral resolution of $\sigma_{\rm LSF} = 0.68$ Å ($\sigma_{v,\rm LSF} = 29.6$ km s⁻¹) in the red arm. For reference, we show the $1\sigma_{v,\rm LSF}$ and $1/2\sigma_{v,\rm LSF}$ on Fig. 3. In total, 89 per cent of our galaxies have estimated intrinsic velocity dispersions less than $\sigma_{v,\rm LSF}$ and 4.6 per cent of our sample were estimated to have intrinsic velocity dispersion less than $\sigma_{v,\rm LSF}/2$.

We correct for the LSF by convolving the emission line by a Gaussian profile with $\sigma_{v,LSF}$ during the fitting procedure in BLOBBY3D. This procedure assumes that the observed emission line is a convolution of two Gaussians. Therefore, the estimated velocity dispersion can be affected by non-Gaussianities in the shape of the LSF, particularly when the velocity dispersion is significantly less than the width of the LSF. However, deviations of the SAMI LSF from a Gaussian profile are minor (van de Sande et al. 2017). Also, 95.4 per cent of our sample were estimated to be $\sigma_{v,0} > \sigma_{v,LSF}/2$, as such the effects of minor systematic differences of the LSF from a Gaussian profile is unlikely to have significant effects on our inferences.

Similarly, the effect of variations in the LSF FWHM is minor for the SAMI Galaxy Survey. The LSF FWHM varied at the ~5 per cent level as a function of fibre, time, and wavelength during the SAMI Galaxy Survey (Scott et al. 2018). For the velocity dispersions values that we estimate, this should result in uncertainties on the level of $\Delta \sigma_v \sim 1 \,\mathrm{km \, s^{-1}}$. As such, the variation of the LSF FWHM is not expected to have any significant effect on the conclusions drawn in this paper.

3.2.3 Estimating the vertical velocity dispersion

Our disc modelling approach calculates a global estimate for the intrinsic LoS velocity dispersion ($\sigma_{v,0} \equiv \sigma_{v,LoS}$). Most studies using IFS observations report $\sigma_{v,LoS}$. However, $\sigma_{v,LoS}$ is a mixture of the radial ($\sigma_{v,R}$), azimuthal ($\sigma_{v,\phi}$), and vertical ($\sigma_{v,z}$) velocity dispersion components.

At any point in the sky, $\sigma_{v,LoS}$ is given by (e.g. equation 27a, Cappellari 2019)

$$\sigma_{v,\text{LoS}}^2 = \left(\sigma_{v,R}^2 \sin^2 \phi + \sigma_{v,\phi}^2 \cos^2 \phi\right) \sin^2 i + \sigma_{v,z}^2 \cos^2 i.$$
(5)

Observed $\sigma_{v,LoS}$ is the luminosity-weighted integral along the LoS. To calculate the average velocity dispersion, we make the following approximations. We assume that the flux is constant across a thin disc with finite radial extent. We also assume spatially constant velocity dispersion components and that $\sigma_{v,\perp}^2 \equiv \sigma_{v,R}^2 \approx \sigma_{v,\phi}^2$ then the average LoS velocity dispersion is given by

$$\bar{\sigma}_{v,\text{LoS}}^2 = \sigma_{v,\perp}^2 \sin^2 i + \sigma_{v,z}^2 \cos^2 i.$$
(6)

Setting $\gamma^2 = \sigma_{v,z}^2 / \sigma_{v,\perp}^2$, and rearranging, then

$$\sigma_{v,\text{LoS}} = \sigma_{v,z} \sqrt{\sin^2 i/\gamma^2 + \cos^2 i}.$$
(7)

The above model predicts changing $\sigma_{\nu,\text{LoS}}$ as a function of *i* if $\gamma \neq 1$. For $\gamma > 1$, $\sigma_{\nu,\text{LoS}}$ increases with increasing *i*, whereas $\sigma_{\nu,\text{LoS}}$ decreases with *i* when $\gamma < 1$.

To estimate γ , we assume that $\sigma_{v,LoS}$ follows a loguniform distribution with mean $\sigma_{v,z,0}$ and log variance τ^2 . The generating function for a single data point $\sigma_{v,z,i}$ is then

$$p(\sigma_{v,\text{LoS},j}|\sigma_{v,z,0},\tau^2,\gamma) \sim \text{lognormal}(\sigma_{v,z,0}\sqrt{\sin^2 i/\gamma^2 + \cos^2 i},\tau^2).$$
(8)

We assume the following priors

$$p(\sigma_{v,z,0}) \sim \text{loguniform}(1, 100)$$
 (9)

$$p(\tau) \sim \text{loguniform}(10^{-3}, 1) \tag{10}$$

 $p(\gamma) \sim \text{loguniform}(0.1, 10).$ (11)

The posterior distribution is then given by

$$p(\sigma_{v,z0}, \tau, \gamma | \mathbf{D}) = p(\sigma_{v,z,0}) p(\tau) p(\gamma) \prod_{j=1}^{N} p(\sigma_{v,\text{LoS},j} | \sigma_{v,z,0}, \tau^{2}, \gamma).$$
(12)



Figure 4. Corner plot (Foreman-Mackey 2016) showing the marginalized (the diagonal) and joint (the off-diagonal) posterior distributions for the parameter estimation for the inclination dependence model. There is evidence for a dependence of $\sigma_{v,LoS}$ on inclination for our sample of galaxies from the SAMI Galaxy Survey. This suggests that the vertical velocity dispersion ($\sigma_{v,z}$) is less than the averaged azimuthal and radial velocity dispersion ($\sigma_{v,\perp}$).

The above formulation assumes independence of the prior distribution between $\sigma_{v,z,0}, \tau, \gamma$, as well as all $\sigma_{v,LoS,j}$. The above posterior distribution can now be sampled using typical techniques. We used EMCEE to sample the posterior distribution (Foreman-Mackey et al. 2013).

We estimate $\gamma = 0.80^{+0.06}_{-0.05}$ as shown in Fig. 4, suggesting that the vertical velocity dispersion is less than the averaged azimuthal and radial components. This analysis was consistent with other approaches that we applied. For example, the bootstrapped Spearmanrank correlation coefficient distribution between the inclination and $\sigma_{v,\text{LoS}}$ was $\rho(i, \sigma_{v,\text{LoS}}) = 0.18^{+0.05}_{-0.05}$, where the uncertainties for the Sperman-rank correlation coefficient is estimated as the 68 per cent shortest credible interval after bootstrap resampling. We also performed the above analysis using uniform priors for $\sigma_{v,z,0}$ and γ with the same ranges, yet we still find $\gamma = 0.80^{+0.06}_{-0.06}$.

Previous studies have suggested that $\sigma_{v,z}/\sigma_{v,R} \sim 0.6$ (section 1.2.2, Glazebrook 2013) for stars. Mean H I gas velocity dispersion was reported up to ~3 times higher for galaxies observed at $i > 60^{\circ}$ compared to $i < 60^{\circ}$ by Leroy et al. (2008), also suggesting that the contribution of $\sigma_{v,R}$ and $\sigma_{v,\phi}$ dominates.

Studies of gas kinematics have typically not reported or found evidence that $\sigma_{v,z}$ is related to the inclination. For example, studies of high z in the KMOS3D Survey have found no significant correlation between the axial ratio q = b/a and $\sigma_{v,LoS}$ (Wisnioski et al. 2015; Übler et al. 2019). However, such a relation may be difficult to identify in high-z galaxies with lower signal-to-noise and spatial resolution.

We estimate the vertical velocity dispersion ($\sigma_{v,z}$) for individual galaxies by inverting equation (7) and using $\gamma = 0.8$. We estimated the Spearman-rank correlation between the inclination and $\sigma_{v,z}$ to be $\rho(i, \sigma_{v,z}) = 0.00^{+0.05}_{-0.05}$ after performing the correction per galaxy,



Figure 5. The relationship between the inclination (*i*) and inferred velocity dispersion estimates. We also show the PDF of the Spearman-rank correlation coefficients using bootstrap resampling (bottom). There is evidence for a weak positive correlation between the LoS velocity dispersion $\sigma_{v,LoS}$ and *i*, whereas the distribution for the vertical velocity dispersion after applying a correction factor yields no relation with *i*.



Figure 6. The distribution of the LoS ($\sigma_{v,LoS}$, blue) and vertical ($\sigma_{v,z}$, red) velocity dispersion for our sample of galaxies from the SAMI Galaxy Survey. The estimated vertical velocity dispersion is adjusted down with respect to $\sigma_{v,LoS}$ by a couple of km s⁻¹ in accordance with the inclination correction described in Section 3.2.3.

suggesting that our analysis appropriately removed the correlation as a function of the inclination angle (see Fig. 5).

Converting from $\sigma_{v,LoS}$ to $\sigma_{v,z}$ adjusts the reported values by a couple of km s⁻¹. The marginalized distributions yield $\sigma_{v,LoS} = 21.1^{+3.9}_{-5.2}$ km s⁻¹ and $\sigma_{v,z} = 18.8^{+3.4}_{-4.8}$ km s⁻¹ (see Fig. 6). Typi-

cal differences are $\sigma_{v,\text{LoS}} - \sigma_{v,z} = 2.4^{+0.9}_{-1.3} \text{ km s}^{-1}$, with the greatest correction being $\sigma_{v,\text{LoS}} - \sigma_{v,z} = 7.9 \text{ km s}^{-1}$.

For the remainder of this paper, we will report the values of $\sigma_{v,z}$. The subsequent analysis and results do not change qualitatively whether we use $\sigma_{v,z}$ or $\sigma_{v,LoS}$, but $\sigma_{v,z}$ is preferred as it is an estimator free from effects from the viewing angle. It is also more appropriate to compare $\sigma_{v,z}$ to theoretical models, as they are typically framed with respect to $\sigma_{v,z}$. We report both values in Appendix A.

We have not applied the inclination correction for galaxies observed in the DYNAMO survey. This is due to finding no significant relation with $\rho(i, \sigma_{v,\text{los}}) = -0.09^{+0.15}_{-0.15}$ for our galaxies from the DYNAMO survey. This suggests that there is no inclination effect to correct for within this sample. It may be that the sample from the DYNAMO survey is too small to infer the inclination effect. In this case, we choose not to apply the inclination effect found from the SAMI Galaxy Survey, as it is still possible that the inferred effect is methodological rather than physical across all galaxies.

3.3 Circular velocity estimates

BLOBBY3D estimates the LoS velocity profile using the empirical model proposed by Courteau (1997):

$$v(r) = v_{\rm c} \frac{(1 + r_{\rm t}/r)^{\beta}}{(1 + (r_{\rm t}/r)^{\gamma})^{1/\gamma}} \sin(i) \cos(\theta) + v_{\rm sys},$$
(13)

where v_c is the asymptotic velocity and v_{sys} is the systemic velocity. *r* is defined by the distance to the kinematic centre. r_t is the turnover radius. β is a shape parameter that controls the gradient for $r > r_t$, where the velocity gradient increases for $\beta < 0$, and decreases when $\beta > 0$. γ is a shape parameter that controls how sharply the velocity profile turns over. *i* is the inclination of the galaxy. Then, θ is the polar angle in the plane of the disc.

We intend to estimate the circular velocity from our inferred parameters. While v_c is a natural choice, it is difficult to get a strong constraint on v_c across our complete sample due to the FoV for the SAMI Galaxy Survey typically extending out to ~1.5 R_e . Instead, we estimate the absolute circular velocity at 2.2 R_e denoted as $v_{2.2}$ following (Bloom et al. 2017a).

For low values of *i*, small differences in the estimated *i* can result in large difference of $v_{2,2}$. Therefore, for low values of *i*, incorrect estimates for the observed ellipticity can result in large changes in our estimates for the inclination. As such, we restrict our calculated values for $v_{2,2}$ to galaxies in the range $i \in [30^\circ, 60^\circ]$ ($e \in [0.13, 0.5]$ assuming a thin disc).

Similarly, galaxies with $R_e < 3.0$ arcsec tended to have very large scatter on their $v_{2,2}$. At these limits, the spatial resolution of our observations are likely playing a role in increasing the scatter in the rotational velocity estimates. A total of 230 galaxies meet the above inclination and R_e criteria. We only reference $v_{2,2}$ for galaxies that meet that inclination for the remainder of this paper.

3.4 Integrated star formation rates

We used the best-fitting SFRs from the GAMA Survey (Gunawardhana et al. 2013; Davies et al. 2016; Driver et al. 2018). The SFRs are estimated using full spectral energy distribution fitting of 21 bands of photometry across the UV, optical, and far-infrared ranges with MAGPHYS (da Cunha, Charlot & Elbaz 2008). MAGPHYS fits the observed photometry using a library that includes stellar spectral and dust emission profiles. In this way, the SFRs are corrected for dust emission. These estimates for the SFR were used instead of the SAMI H α luminosity maps as there are known aperture affects given the limited FoV of the SAMI instrument (appendix A, Medling et al. 2018).

For the galaxies from the DYNAMO survey, we used the SFR values reported by Green et al. (2014). SFRs were estimated using the H α luminosity estimated from their observations. The SFR estimate includes a dust correction using the Balmer decrement from the ratio between their measured H α and H β measurements. The SFR was then calculated using the dust-corrected H α luminosity maps that were converted to SFR maps using the Kennicutt, Tamblyn & Congdon (1994) conversion assuming a Chabrier (2003) IMF.

3.5 Integrated H I gas measurements

Follow-up 21 cm observations of SAMI galaxies were obtained as part of the SAMI-HI survey, carried out with the Arecibo radio telescope (Catinella et al., in preparation). Observations and data reduction were analogous to those of the xGASS survey (Catinella et al. 2018), with the only difference that these were not gas fraction-limited observations. We observed each galaxy until detected, but moved to another target if there was no hint of HI signal within the first 20–25 min of on-source integration.

H I emission-line spectra were obtained for 153 galaxies with these dedicated follow-up observations; on-source integration times ranged between 2 and 50 min, with an average of 15 min. Together with an additional 143 good H I detections (i.e. classified as detection code '1') in the Arecibo Legacy Fast ALFA (Giovanelli et al. 2005; Haynes et al. 2018) survey, SAMI–H I includes global H I spectra for 296 SAMI galaxies from the SAMI Galaxy Survey catalogue. A total of 95 galaxies overlap with our sample selection from the SAMI Galaxy Survey.

4 RESULTS

4.1 Low gas velocity dispersion in the SAMI Galaxy Survey

We find vertical velocity dispersions lower than previously reported for studies of the gas kinematics in the SAMI Galaxy Survey. The median vertical velocity dispersion is $\sigma_{v,z} = 18.8 \text{ km s}^{-1}$ for our sample as shown in Fig. 6. The 68th shortest credible interval is [14.1, 22.1] km s⁻¹ and the 95th shortest credible interval is [11.4, 30.0] km s⁻¹. The maximum inferred vertical velocity dispersion for a single galaxy is $\sigma_{v,z} = 51 \text{ km s}^{-1}$. We now compare this to two other studies of the gas kinematics of galaxies from the SAMI Galaxy Survey by Zhou et al. (2017) and Johnson et al. (2018).

Analysing eight star-forming galaxies in the SAMI Galaxy Survey, Zhou et al. (2017) found that seven of eight galaxies had $\sigma_{gas} \in [20, 31] \text{ km s}^{-1}$. Their remaining galaxy (GAMA 508421) was reported as $\sigma_{gas} = 87 \pm 44 \text{ km s}^{-1}$. GAMA 508421 exhibits a high circular velocity in the outskirts of the SAMI FoV ($v \sim 130 \text{ km s}^{-1}$) and a clear centralized peak in velocity dispersion that is typical of beam smearing affected galaxies. Our estimate for GAMA 508421 is $\sigma_{v,z} = 22 \text{ km s}^{-1}$. As such, we suspect that the reported velocity dispersion for GAMA 508421 is greater than its intrinsic velocity dispersion.

The discrepancy between Zhou et al. (2017) and our estimates, particularly with GAMA 508421, is most likely due to the different beam smearing corrections. Zhou et al. (2017) report the flux-weighted mean velocity dispersion using spaxels where $\sigma_v > 2v_{\text{grad}}$.

 v_{grad} is an estimate for the local velocity gradient using adjacent spaxels defined as (Varidel et al. 2016)

$$v_{\text{grad}}(x, y) = \sqrt{(v(x+1) - v(x-1))^2 + (v(y+1) - v(y-1))^2}.$$
(14)

See section 5.1.1 by Varidel et al. (2019) for a revised calculation of the velocity gradient using a finite-difference scheme.

The approach used by Zhou et al. (2017) usually removes the centre of the galaxies, where the velocity gradient is steepest. This approach results in a significant downward correction compared to the uncorrected velocity dispersion estimates. However, the outskirts of galaxies can still be affected by beam smearing. Also, it is possible that the centre of the galaxy may be effected by beam smearing, yet not reach the $\sigma_v > 2v_{\text{grad}}$ criteria, which is likely to have occurred in the case of GAMA 508421. The approach of Zhou et al. (2017) was also shown previously to overestimate the intrinsic velocity dispersion in toy models (section 5.1.1., Varidel et al. 2019).

Another study of a sample of 274 star-forming galaxies from the SAMI Galaxy Survey was performed by Johnson et al. (2018). They removed galaxies with $M_* > 8 \times 10^{10} \text{ M}_{\odot}$ and Sérsic index of n > 2. They also removed galaxies that they deem to be spatially unresolved or have kinematic uncertainties greater than 30 per cent. While they do not provide summary statistics for their inferred velocity dispersion values from the SAMI Galaxy Survey, their plots show a typical range of $\sigma_0 \in [20, 60] \text{ km s}^{-1}$, plus one galaxy at $\sigma_0 \sim 90 \text{ km s}^{-1}$. This is slightly above our range of velocity dispersions.

To estimate the intrinsic velocity dispersion, Johnson et al. (2018) calculated the median velocity dispersion across the kinematic maps or at the outskirts of their galaxy. They then apply a further correction on their estimated velocity dispersion using a lookup table of toy galaxies that have been constructed with beam smearing effects. The slight difference between our studies may be driven solely by their choice of using a single FWHM estimate for the PSF rather than the Moffat profile used in this paper. Also, increased scatter may occur in their estimator due to being affected by low signal-to-noise spaxels in the outskirts of the galaxies.

4.2 Correlation of global velocity dispersion and integrated star formation rate

Correlation analysis between the global velocity dispersion and several global galaxy properties from the SAMI Galaxy Survey reveals that $\sigma_{v,z}$ has the greatest positive correlation with SFR measures (Fig. 7). We estimate the Spearman-rank correlation between the SFR and $\sigma_{v,z}$ to be ρ (SFR, $\sigma_{v,z}$) = 0.44^{+0.05}_{-0.05}. We control for several factors in order to investigate this relationship further.

The correlation between $\sigma_{v,z}$ and SFR increases when accounting for the galaxy size. To do this, we estimate the average SFR surface density, $\Sigma_{SFR} = SFR/\pi R_e^2$, where R_e is the effective radius. The Spearman-rank correlation is then $\rho(\Sigma_{SFR}, \sigma_{v,z}) = 0.54^{+0.04}_{-0.04}$. Velocity dispersion is expected to increase with star formation rate surface density assuming that star formation feedback processes are acting as a driver of turbulence (e.g. Ostriker & Shetty 2011; Faucher-Giguère et al. 2013). As such, this does provide support that star formation feedback processes is acting as a driver of turbulence within this sample of galaxies.

Fig. 7 also shows a positive correlation between $\sigma_{\nu,z}$ and integrated stellar mass (M_{*}), H I gas mass (M_{H I}), as well as the sum of M_{*} and M_{H I}. Interestingly, there is a suggestion that M_{H I} is slightly



Figure 7. Comparing global intrinsic vertical velocity dispersion ($\sigma_{v,z}$) to global properties for galaxies from the SAMI Galaxy Survey. We show the relation of $\sigma_{v,z}$ with measures of mass (top), star formation rate (middle), and rotational velocity (bottom), respectively. The red points indicate the galaxies with observed integrated H_I masses. The Spearman-rank correlation coefficients are shown at the top of each plot, with brackets indicating the correlation coefficient for galaxies with measured H_I masses. The uncertainties for the Spearman-rank correlation coefficients are estimated as the 68 per cent shortest credible interval from 10⁴ bootstrapped samples. We find significant positive correlations with measures of mass, star formation rate, and rotational velocity. The greatest positive correlation we find is with star formation rate surface density (Σ_{SFR}).

more correlated than M_* with $\sigma_{v,z}$, although the uncertainties are wide enough that we cannot confirm that is the case. SFR is well known to be correlated with M_* , which adds a further complication in determining the relation between $\sigma_{v,z}$ and SFR.

To account for the SFR–M_{*} relation, we calculated the specific SFR (sSFR = SFR/M_{*}) and Δ MS. Δ MS is calculated as the log difference between the SFR and the star-forming main-sequence relation as proposed by Renzini & Peng (2015). We find that the correlation between $\sigma_{v,z}$ and SFR decreased after accounting for stellar mass. This suggests that the relation between $\sigma_{v,z}$ and SFR is a combination of both SFR and stellar mass related quantities.

Despite the correlation between $\sigma_{v,z}$ and SFR estimators, the absolute change in $\sigma_{v,z}$ as a function of SFR remains slight across the dynamic range of SFR $\in [10^{-3}, 10] \, M_{\odot} \, yr^{-1}$. We report the change in velocity dispersion in five SFR bins in Table 1. The change in mean velocity dispersion between the end bins from SFR $= 0.029 \, M_{\odot} \, yr^{-1}$ to SFR $= 2.4 \, M_{\odot} \, yr^{-1}$ is only 6.41 km s⁻¹. A similarly shallow gradient was found by Johnson et al. (2018) using data from the SAMI Galaxy Survey.

Galaxies are often kinematically classified as either rotationally or turbulence dominated by comparing the ratio of rotational and random velocities (v/σ) . In a similar vain to such analysis, we also investigated the relation between $\sigma_{v,z}$ and rotational velocity. $\sigma_{v,z}$ is shown compared to the rotational velocity measures using BLOBBY3D ($v_{2,2}$) as outlined in Section 3.3 and using the Tully–Fisher relation ($v_{2,2,tf}$, Bloom et al. 2017b).

We find a positive correlation between $\sigma_{v,z}$ and the rotational velocity estimators. This is to be expected as rotational velocity is also correlated with stellar mass. To control for that effect, we calculated the ratio between $v_{2,2}$ and $v_{2,2,tf}$. We then find a negative correlation between $\sigma_{v,z}$ and $v_{2,2/v_{2,2,tf}}$. As such, we observe that galaxies exhibit greater rotation than their mass predicts when $\sigma_{v,z}$ is lesser, and lesser rotation when $\sigma_{v,z}$ is greater.

4.3 Comparisons with other surveys

In this section, we aim to describe our results from the SAMI Galaxy Survey in the context of other studies. In Table 1 and Fig. 8, we show comparisons of velocity dispersion compared to SFR. The data are shown in four groups of galaxies: low-*z* measured using H α (Epinat et al. 2008; Moiseev et al. 2015), low-*z* measured using HI (Leroy et al. 2008; Walter et al. 2008; Ianjamasimanana et al. 2012; Stilp et al. 2013), High-*z* analogues from Varidel et al. (2016) plus the galaxies that we re-analysed from the DYNAMO sample, and high-*z* galaxies at $z \gtrsim 1$ (Cresci et al. 2009; Epinat et al. 2009; Law et al. 2009; Jones et al. 2010; Wisnioski et al. 2011; Di Teodoro

Table 1. Comparing summary statistics of the vertical velocity dispersion in other samples compared to those in this work. Each sample was split into five bins of equalpercentile widths. We show the mean $(\bar{\sigma}_{v,z})$, standard deviation $(\Delta \sigma_{v,z})$, the standard error $(\Delta \bar{\sigma}_{v,z})$, median [med $(\sigma_{v,z})$], and bootstrap resampled standard deviation of the median [$\Delta \text{med}(\sigma_{v,z})$]. The groups of galaxies are as follows: Low-*z* (H α ; Epinat et al. 2008; Moiseev et al. 2015), H I surveys where 15 km s⁻¹ has been added in-quadrature (Leroy et al. 2008; Walter et al. 2008; Ianjamasimanana et al. 2012; Stilp et al. 2013), high-*z* analogues from Varidel et al. (2016) plus the re-analysed galaxies from the DYNAMO survey, plus high-*z* (H α ; Cresci et al. 2009; Epinat et al. 2009; Jones et al. 2010; Wisnioski et al. 2011; Di Teodoro, Fraternali & Miller 2016; Johnson et al. 2018).

Group	Bin	$SFR~(M_\odot~yr^{-1})$	$\bar{\sigma}_{v,z} (\mathrm{km} \mathrm{s}^{-1})$	$\Delta \sigma_{v,z} (\mathrm{km} \mathrm{s}^{-1})$	$\Delta \bar{\sigma}_{v,z} (\mathrm{km} \mathrm{s}^{-1})$	$\operatorname{med}(\sigma_{v,z}) (\mathrm{km}\mathrm{s}^{-1})$	$\Delta \mathrm{med}(\sigma_{v,z}) (\mathrm{km}\mathrm{s}^{-1})$
SAMI (Η α)	1	0.029	17.12	3.21	0.39	17.13	0.29
	2	0.11	18.54	3.99	0.49	18.31	0.41
	3	0.25	18.79	4.34	0.53	18.52	0.43
	4	0.57	21.07	6.47	0.79	19.72	0.71
	5	2.4	23.54	5.35	0.65	23.54	0.64
Low- z (H α)	1	0.0047	19.46	2.89	0.43	18.84	0.72
	2	0.046	20.77	4.33	0.65	19.21	0.41
	3	0.18	20.57	3.86	0.58	19.21	0.6
	4	0.37	21.66	4.55	0.68	19.85	0.44
	5	1.0	23.5	7.0	1.0	21.21	0.81
Low- <i>z</i> (H I)	1	0.0014	16.95	0.55	0.18	16.86	0.15
	2	0.005	17.39	0.64	0.20	17.44	0.25
	3	0.066	18.65	2.98	0.99	17.81	0.6
	4	0.58	19.18	1.36	0.43	18.78	0.57
	5	2.2	20.82	2.58	0.82	19.9	1.4
High z	1	0.96	27.0	3.2	1.1	26.23	0.94
Analogues $(H\alpha)$	2	3.2	39.4	12.6	4.4	40.0	4.9
	3	9.1	40.7	14.3	5.0	41.2	7.8
	4	17	43.0	15.2	5.4	42.9	7.6
	5	27	55.9	15.6	5.2	54.8	5.4
High z (H α)	1	3.4	44.0	20.5	1.6	39.8	1.9
	2	6.4	45.8	18.2	1.5	43.1	1.2
	3	10	44.3	20.3	1.6	42.8	3.2
	4	20	48.3	20.2	1.6	45.0	1.5
	5	82	53.2	20.0	1.6	51.0	2.6

et al. 2016; Johnson et al. 2018). Table 2 also outlines qualitative ranges for the galaxy parameters for galaxies at low *z* measured using the H α emission line, including other studies of the SAMI and DYNAMO samples.

The comparative data sets have been measured using both ionized and neutral gas. For ionized gas, there are two additional contributions to the velocity dispersion. One is the thermal broadening of $\sigma_{\rm thermal} \sim 9 \,\rm km \, s^{-1}$, corresponding to the typical temperature of an H II region. There is also a contribution from the expansion speed of the H II region. Studies of the expansions speed reveal $\sigma_{\rm expand} \sim 10 \,\rm km \, s^{-1}$ for small regions, up to $\sigma_{\rm expand} \sim 13-17 \,\rm km \, s^{-1}$ for larger regions (Chu & Kennicutt 1994).

Given the contributions of σ_{thermal} and σ_{expand} to the observed ionized gas kinematics, we perform several adjustments to the comparative velocity dispersion estimates. For ionized gas estimates, we remove any corrections for the additional contributions. For H I studies, we assume a nominal contribution due to these effects of 15 km s⁻¹, which we add in quadrature to the published velocity dispersion estimates. We note that in other studies, 15 km s⁻¹ has been subtracted in quadrature from the ionized gas measurements for comparisons between different studies. We prefer the alternative as 15 per cent of our galaxies have $\sigma_{v,z} < 15 \text{ km s}^{-1}$.

4.3.1 Comparison with surveys at low z

The SAMI Galaxy Survey has similar selection criteria to the Mapping Nearby Galaxies at Apache Point Observatory (MaNGA;

Bundy et al. 2015) survey in terms of fundamental galaxy properties (see Table 2). Our data have similar ranges in redshifts, stellar mass, and SFR. As such, we would naively expect the gas turbulence within our sample to be similar to the MaNGA survey estimates.

We find systematically lower velocity dispersions than those estimated by Yu et al. (2019). They estimated mean velocity dispersions of $\sigma \in [20, 50] \text{ km s}^{-1}$ across various galaxy property ranges (Fig. 6, Yu et al. 2019). Specifically for SFR versus velocity dispersion they found mean $\sigma \in [30, 50] \text{ km s}^{-1}$ across four bins in the range SFR $\in [10^{-2}, 10] \text{ M}_{\odot} \text{ yr}^{-1}$, whereas we estimate mean $\bar{\sigma}_{v,z} \in [17, 24] \text{ km s}^{-1}$ across five bins of SFR $\in [10^{-3}, 10]$.

Yu et al. (2019) also reported galaxies with velocity dispersion of $\sigma_v \gtrsim 50 \,\mathrm{km \, s^{-1}}$ up to $\sigma_v \sim 130 \,\mathrm{km \, s^{-1}}$. This is similar to σ_v estimates for galaxies at high redshift (see high-*z* galaxies, Table 1). However, we see very little evidence for a significant fraction of galaxies with $\sigma_v \gtrsim 50 \,\mathrm{km \, s^{-1}}$.

The spectral resolution of $\sigma_{\rm LSF} \in [50, 80] \,\rm km \, s^{-1}$ (Bundy et al. 2015; Yu et al. 2019) may be an issue for MaNGA. The variability in the MaNGA spectral resolution could correspond to a large scatter in their estimated velocity dispersion, which may explain their upper limit of $\sigma_v \sim 100 \,\rm km \, s^{-1}$. We also show that the velocity dispersion is significantly less than their spectral resolution, thus their assumptions regarding the LSF will be important.

Instead, our results are closer to the velocity dispersion estimates found in the Gassendi HAlpha survey of SPirals (GHASP; Epinat et al. 2008), where their galaxies overlap in SFR. We can see in Fig. 8 that our samples match well with the work of Epinat et al. (2008)



Figure 8. Comparison of the SFR–velocity dispersion (σ_v) relation compared to others surveys in the literature. The sets of galaxies that constitute each subplot are the same as outlined in Table 1. We find the SFR– σ_v relation increases slightly across the range SFR $\in [10^{-3}, 1] M_{\odot} \text{ yr}^{-1}$, then turns up significantly at SFR $\gtrsim 1 M_{\odot} \text{ yr}^{-1}$. This relation is approximately consistent across all surveys.

Table 2. Qualitative ranges of galaxy parameters for low-z samples in the literature, where gas kinematics were estimated using the H α emission line.

Sample	Z.	$log_{10}(M_{*}/M_{\odot})$	$\log_{10}(SFR/M_{\odot} yr^{-1})$	$\sigma_v (\mathrm{kms^{-1}})$
SAMI (this work)	[0.005, 0.08]	[7.5, 11]	[-3, 1]	[10, 60]
SAMI (Johnson et al. 2018)	< 0.1	[7.5, 11]	[-3, 1]	[20, 90]
SAMI (Zhou et al. 2017)	< 0.1	[9.8, 10.8]	_	[20, 90]
DYNAMO (this work)	[0.06, 0.15]	[9, 11]	[-1, 2]	[10, 80]
DYNAMO (Green et al. 2014)	[0.06, 0.15]	[9, 11]	[-1, 2]	[10, 90]
GHASP (Epinat et al. 2008)	~ 0.01	_	[-3, 1]	[15, 30]
Moiseev et al. (2015)	< 90 Mpc	_	[-3, 1]	[15, 40]
Varidel et al. (2016)	[0.01, 0.04]	[10.5, 11]	[1, 1.6]	[20, 50]
MaNGA (Yu et al. 2019)	[0.01, 0.15]	[8.5, 11.5]	[-2, 1]	[10, 130]

both in terms of mean velocity dispersion and gradient as a function of SFR. We only disagree slightly in terms of the intrinsic scatter, which could be sample selection, methodology, or signal-to-noise dependent.

We highlight that Epinat et al. (2008) estimated their velocity dispersion using the residuals in spatially resolved mean velocity compared to a rotational velocity model. As such, their measurements are fundamentally different and should not be affected by σ_{thermal} and σ_{expand} . So, we added 15 km s⁻¹ in quadrature to their published velocity dispersion estimates for comparison purposes.

Our results are also qualitatively similar to those published by Moiseev et al. (2015), who studied a sample of nearby dwarf galaxies. Their results agree with the higher end of our velocity dispersion estimates, although there is still an offset in the mean velocity dispersion. We note that Moiseev et al. (2015) do not explicitly correct for beam smearing, but due to studying nearby galaxies at <90 Mpc, the effects of beam smearing should be minimal.

Combining the results of Moiseev et al. (2015) and Epinat et al. (2008), we find differences of the mean and median velocity dispersion estimates compared to our sample of $\sim 1-3 \text{ km s}^{-1}$ (see Table 1), where our results were systematically lower. The difference of $\sim 2 \text{ km s}^{-1}$ could be explained due to calculating $\sigma_{v,z}$ rather than $\sigma_{v,LoS}$, which resulted in a downward shift in our velocity dispersion estimates by $\sim 2 \text{ km s}^{-1}$ as described in Section 3.2.3.

We find little difference in the intrinsic scatter between our sample and the combined samples of Moiseev et al. (2015) and Epinat et al. (2008). Calculating the 1 σ standard deviation for the sample ($\Delta \sigma_{v,z}$), sample mean ($\Delta \bar{\sigma}_{v,z}$), and median ($\Delta \text{med}(\sigma_{v,z})$, we find that all variance estimates were of similar magnitude (see Table 1). As such, we conclude that our results are approximately consistent with the analyses of Moiseev et al. (2015) and Epinat et al. (2008) at low *z* using ionized gas, albeit with different selection and methodologies in inferring the intrinsic velocity dispersion. The only exception in inferred velocity dispersions at low *z* using the ionized gas is the results of Yu et al. (2019) using MaNGA data where we estimate systematically lower σ_v .

Comparisons to the HI observations suggest that we get the same approximately flat SFR- σ_v relation across the range SFR \in [10⁻³, 10] M_{\odot} yr⁻¹. While there are only slight differences between the mean velocity dispersion of \sim 1–4 km s⁻¹ across varying SFR ranges, it is important to reiterate that the HI results have 15 km s⁻¹ added in quadrature, which is the typical difference between HI and H α estimates for the velocity dispersion. The varying contributions of σ_{thermal} and σ_{expand} may cause a larger scatter than the neutral hydrogen estimates.

4.3.2 Comparisons with surveys at high-z and high-z analogues

We now compare our results to those at high-*z* and high-*z* analogues. The data sets included are from the DYNAMO survey, which we have re-analysed using BLOBBY3D. We also include the beamsmearing corrected estimates denoted as $\sigma_{m,uni,v_g=0}$ from Varidel et al. (2016). These samples are of galaxies at low *z* with SFR $\gtrsim 1 \text{ M}_{\odot} \text{ yr}^{-1}$, which are similar to galaxies at high *z* (see Table 1). As such, high-*z* analogues are likely to have similar properties to our galaxy sample at similar SFR.

Our re-analysis of the galaxies from the DYNAMO survey find results consistent with Green et al. (2014). The difference between our results and those of Green et al. (2014) are $\sigma_{v,z}-\sigma_{v,green} =$ $0.0^{+4.9}_{-6.5}$ km s⁻¹. Follow-up studies of galaxies from the DYNAMO survey have also found similar results including re-analysis using alternative beam smearing corrections (Bekiaris et al. 2016) and observations using adaptive optics (Oliva-Altamirano et al. 2018).

There is a slight increase in σ_v when comparing SAMI with the high-z analogues at overlapping SFR. At SFR $\sim 3 M_{\odot} \text{ yr}^{-1}$, we estimate $\bar{\sigma}_{v,\text{SAMI}} = 23.54 \pm 0.65 \text{ km s}^{-1}$ compared to $\bar{\sigma}_{v,\text{HzA}} =$ $27.0 \pm 1.1 \text{ km s}^{-1}$ at SFR $\sim 2.4 \text{ M}_{\odot} \text{ yr}^{-1}$ and $\bar{\sigma}_{v,\text{HzA}} = 39.4 \pm 4.4 \text{ km s}^{-1}$ at SFR $\sim 3.2 \text{ M}_{\odot} \text{ yr}^{-1}$ for the high-z analogues. The highest velocity dispersions are primarily from the DYNAMO survey. We note that while BLOBBY3D was applied to both samples, the PSF for DYNAMO was assumed to be a Gaussian profile compared to a Moffat profile for the SAMI Galaxy Survey. This may result in an increased beam smearing correction in the SAMI Galaxy Survey compared to the DYNAMO survey. Also, the inclination correction was only applied to SAMI, which resulted in an $\sim 2 \text{ km s}^{-1}$ subtraction to the initially inferred velocity dispersion from BLOBBY3D. As such, a difference of $\sim 10 \text{ km s}^{-1}$ may not be significant given limitations of comparing the two samples.

The high-*z* analogues extend the trend of increasing σ_v with SFR (see Fig. 8). This trend starts to increase within the sample from SAMI Galaxy Survey at SFR $\geq 1 \, M_{\odot} \, yr^{-1}$. Expanding the SFR range up to SFR $\sim 100 \, M_{\odot} \, yr^{-1}$ using the high-*z* analogues, we see that trend increases dramatically with σ_v up to 80 km s⁻¹ in the range SFR $\in [10, 100] \, M_{\odot} \, yr^{-1}$, which is qualitatively consistent with samples at high *z*.

The high-*z* galaxies exhibit a wide range of $\sigma_v \in [10, 150] \text{ km s}^{-1}$. Some of this extent is likely to be driven by lower signal to noise at higher redshift. Furthermore, systematic biases such as beam smearing effects, which act to increase σ_v , will be greater due to the lower spatial resolution. Instead, the high-*z* galaxies still exhibit similar σ_v as the high-*z* analogues when studied as a group.

The high-z galaxies still exhibit a trend of increasing velocity dispersion as a function of SFR. There is a change from $\sigma_v \sim 40$ to $\sim 50 \text{ km s}^{-1}$ for SFR of $3-82 \text{ M}_{\odot} \text{ yr}^{-1}$ (see Table 1). We estimated the correlation to be $\rho(\text{SFR}, \sigma_v) = 0.17^{+0.03}_{-0.04}$. This is a weaker correlation between SFR and σ_v than observed in low-z galaxies. Lesser correlation is likely linked to increased scatter for observations of galaxies at high z. The increase in scatter may be driven by signal-to-noise, beam smearing effects due to lower spatial resolution, or a change in the physical drivers of gas turbulence at high z.

There is evidence for increased σ_v at high *z* compared to the high-*z* analogues at similar SFRs. In Table 1, we show binned estimators for dynamic ranges of SFR \in [3, 30] M_{\odot} yr⁻¹ for these two samples. σ_v is ~5 km s⁻¹ higher at similar SFRs for the high-*z* galaxies compared to the high-*z* analogues.

5 THE DRIVERS OF TURBULENCE WITHIN LOW-z GALAXIES

Turbulence in the ISM is expected to dissipate on the order of the disc crossing time (Mac Low et al. 1998; Stone et al. 1998). Thus, an ongoing energy source is required to maintain supersonic gas turbulence across epochs. Two proposed drivers are star formation feedback process and gravity-driven turbulence.

5.1 Star formation feedback-driven turbulence

Star formation feedback processes inject momentum into the ISM through several mechanisms. These mechanisms include supernova, stellar winds, expanding H II regions, and radiation pressure from highly dense star clusters. Therefore, there has been a claim that star formation feedback processes could provide an ongoing source of energy for the supersonic turbulence in the ISM.

Observational studies have routinely found that there is a positive correlation between global σ_v and SFR, which has been used as evidence to support star formation feedback processes as a driver of turbulence (Green et al. 2010, 2014; Moiseev et al. 2015; Johnson et al. 2018; Übler et al. 2019; Yu et al. 2019). In Section 4.2, we showed that this correlation exists in our sample of galaxies. We also showed that this correlation extends to higher SFR when connecting our sample to other galaxy surveys.

The relationship between SFR and σ_v has also been considered in theoretical and computational studies. Typically, the energy contribution from supernovae is considered to dominate, and therefore has been the primary focus of most of these studies. The momentum injection per mass of stars is often assumed to be on the order of $\langle p_*/m_* \rangle = 3000 \,\mathrm{km \, s^{-1}}$. Incorporating this momentum injection into theoretical models results in assuming that the rate of momentum injection is proportional to the SFR surface density, thus $\dot{P} \propto \langle p_*/m_* \rangle \Sigma_{\rm SFR}$ (e.g. Ostriker & Shetty 2011; Faucher-Giguère et al. 2013; Krumholz et al. 2018). Therefore, we expect the velocity dispersion to be positively correlated with SFR surface density, if star formation feedback processes is playing a role in driving turbulence in the ISM.

We showed in Section 4.2 that $\sigma_{v,z}$ has a strong positive correlation with the galaxy-averaged SFR surface density. This is consistent with other analyses of the SFR density and velocity dispersion (e.g. Lehnert et al. 2009; Übler et al. 2019; Yu et al. 2019). In some cases, this has been used as evidence for star formation feedback processes acting as a primary driver of turbulence (Lehnert et al. 2009, 2013). Yet if star formation feedback processes are acting as a driver of turbulence, we should expect that the localized Σ_{SFR} and σ_v are correlated, yet some analyses have found this relation (Lehnert et al. 2009, 2013), and other studies have found a weak or statistically insignificant relation between these localized properties (Genzel et al. 2011; Varidel et al. 2016; Zhou et al. 2017; Übler et al. 2019). Another approach to compare the observed velocity dispersion to the SFR is to construct a bottom–up approach whereby Σ_{SFR} is modelled on the local scale and then integrated across the disc to estimate SFR.

To estimate Σ_{SFR} as a function of galaxy properties, it is first noted that the SFR surface density is a function of the star-forming molecular gas fraction (f_{sf}) of the gas surface density (Σ_{gas}), which is then converted to stars at an SFR efficiency per free-fall time (ϵ_{ff}). Following Krumholz et al. (2018), this can be written as

$$\Sigma_{\rm SFR} = \frac{\epsilon_{\rm ff}}{t_{\rm ff}} f_{\rm sf} \Sigma_{\rm gas},\tag{15}$$

where the remaining undefined quantity is the free-fall time ($t_{\rm ff}$). This can then be incorporated into models to make predictions for the velocity dispersion.

One approach is to assume that the star formation law is retained on the subgalactic scale. This assumes that $\epsilon_{\rm ff}$ is approximately constant across the galaxy, which is broadly in agreement with the literature (Krumholz & Tan 2007; Krumholz et al. 2012; Federrath 2013; Salim et al. 2015; Krumholz et al. 2019). While some studies have found evidence for varying $\epsilon_{\rm ff}$ as a function of galaxy properties (Hirota et al. 2018; Utomo et al. 2018), the results and implications for the value of $\epsilon_{\rm ff}$ remains in dispute. Furthemore, studies using the above approximation have found that $\sigma_{v,z} \lesssim 25 \,\rm km \, s^{-1}$, with little variation of $\sigma_{v,z}$ as a function of SFR (Ostriker & Shetty 2011; Krumholz et al. 2018). As noted in the above samples, there is a large population of galaxies with $\sigma_{v,z} \gtrsim 25 \,\rm km \, s^{-1}$, particularly at high redshifts. As such, it is unlikely that this model is able to explain the full range of observed $\sigma_{v,z}$. Furthermore, such models Another approach is to assume that $\epsilon_{\rm ff}$ can vary as a function of galaxy properties. One such approach was developed by Faucher-Giguère et al. (2013), which assumes that the Toomre stability criteria Q self-regulates to 1. In their model, when Q < 1 the rate of constructing GMCs increases, thus increasing star formation efficiency, driving Q upwards to 1. When Q > 1 the rate of GMC construction is limited and thus star formation slows, leading to Q decreasing to 1. The Faucher-Giguère et al. (2013) predicts that $\epsilon_{\rm ff}$ increases with molecular gas content of the galaxy, leading to a correlation between SFR and velocity dispersion, thus potentially providing an explanation for the SFR– σ_v relation. Hereafter, we will refer to this model as 'No Transport, Fixed Q' and use the analytical model proposed by Krumholz et al. (2018) for comparison in the following sections.

5.2 Gravity-driven turbulence

An alternative to star formation feedback processes is driving of turbulence due to gravitational mechanisms. In such models, the gravitational potential energy of the gas is converted to kinetic energy, thus driving the turbulence in the ISM. Several mechanisms for this to occur are via accretion on to the disc, accretion through the disc, gravitational instabilities in the disc, or gravitational interactions between components of the disc.

During the initial formation of the disc, there is evidence that accretion on to the disc can cause the high levels of gas turbulence. However, this can only be sustained on the order of the accretion time (Aumer et al. 2010; Elmegreen & Burkert 2010). After initial disc formation, the effect of accretion on to the disc is unlikely to have a significant contribution on the gas turbulence (Hopkins et al. 2013).

Instead, it has been shown that the supersonic turbulence initially set in the ISM during galaxy formation will quickly approach a steady-state solution (Krumholz & Burkert 2010). Such a steadystate solution can be found where the sole driving force is due to the accretion of gas through the disc balanced by the loss of turbulence primarily by shocks. This yields prescriptions for radial models of the gas surface density and $\sigma_{v,z}$. Making simplifying assumptions whereby the entire ISM is assumed to be a single star-forming region, and integrating the models over the radial extent of the disc, they derive a relationship that simplifies to SFR $\propto \sigma_{v,z}$, assuming other disc parameters are constant.

The above model is an instantaneous steady state solution, which is a function of the gas accretion rate and energy loss at the time. As the gas accretion rate has decreased over epochs, this model predicts lower gas turbulence in the ISM of galaxies at low z. In Section 4.3.2, we highlighted that velocity dispersions were $\sim 5 \text{ km s}^{-1}$ higher in the high-z sample compared to the high-z analogues sample at similar SFR. This is consistent with the velocity dispersion decreasing as a function of decreasing gas accretion rate over time. Numerous other studies have also found that gas turbulence increases as a function of z (Kassin et al. 2012; Wisnioski et al. 2015; Johnson et al. 2018; Übler et al. 2019).

5.3 Combining star formation feedback and gravity-driven turbulence

Krumholz et al. (2018) recently pointed out that star formation feedback processes can be added as an extra source of energy to the

transport equation derived in Krumholz & Burkert (2010). Similar to the previously mentioned models for star formation feedback processes, they only assume the contribution of supernovae on the gas turbulence.

Their full 'Transport + Feedback' model gives a SFR– $\sigma_{v,z}$ relation of the form

$$SFR = \frac{2}{1+\beta} \frac{\phi_a f_{sf}}{\pi GQ} f_{g,Q} v_c^2 \sigma_{v,z} \\ \times \max\left[\sqrt{\frac{2(1+\beta)}{3f_{g,P}}} \phi_{mp} \frac{8\epsilon_{ff} f_{g,Q}}{Q}, \frac{t_{orb,out}}{t_{sf,max}}\right],$$
(16)

where $f_{\rm sf}$ is the fraction of the gas in the molecular star-forming phase. $f_{g,P}$ is the fractional contribution of the gas to the selfgravitation pressure at the mid-plane. $f_{g,Q}$ is the fractional gas contribution to the toomre-Q parameter. β describes the slope of the rotation curve ($\beta = d \ln v_c / d \ln r$). $t_{\rm sf,max}$ corresponds to the maximum star formation time-scale. $t_{\rm orb,out}$ corresponds to the orbital period at the edge of the star-forming dominated disc. ϕ_a is a constant that accounts for an offset due to observing global rather than local properties, with $\phi_a = 1$ for local galaxies. $\phi_{\rm mp} = 1.4$ corresponds to the assumed ratio of total pressure compared to turbulent pressure at the mid-plane.

This model results in a SFR– $\sigma_{v,z}$ relation with a floor at 15 km s⁻¹ $\lesssim \sigma_{v,z} \lesssim 25$ km s⁻¹ (including the expansion and thermal contributions) for the lower SFR region, thus reproducing gas turbulence that is consistent with the 'No Transport, Fixed $\epsilon_{\rm ff}$ ' model. The SFR– $\sigma_{v,z}$ relation then transitions to SFR $\propto \sigma_{v,z}$ for higher SFR, consistent with the 'No Feedback' model.

Another important contribution of Krumholz et al. (2018) is that after deriving the transport equation, they can use it to find the steady-state solutions making various assumptions. The above model assumes that there is a contribution of star formation driven turbulence ($\sigma_{v,sf}$) to the total turbulence ($\sigma_{v,z}$), where

$$\sigma_{\nu,\text{sf}} = \frac{4 f_{\text{sf}} \epsilon_{\text{ff}} \langle p_*/m_* \rangle}{\sqrt{3 f_{g,P}} \pi \eta \phi_{\text{mp}} \phi_Q \phi_{\text{nt}}^{3/2}} \times \max\left[1, \sqrt{\frac{3 f_{g,P}}{8(1+\beta)}} \frac{Q_{\text{min}} \phi_{\text{mp}}}{4 f_{g,Q} \epsilon_{\text{ff}}} \frac{t_{\text{orb}}}{t_{\text{sf, max}}}\right].$$
(17)

Here, $\eta = 1.5$ is a scaling parameter for the dissipation rate. $\phi_{\rm mp} = 1.4$ is the ratio of total to turbulent pressure at the mid-plane. $\phi_Q = 2$ is the gas to stellar Q plus one. By setting $\sigma_{v,\rm sf} = 0$, Krumholz et al. (2018) derive the 'No Feedback' model. In that case, the disc must remain stable, such that Q = 1.

Krumholz et al. (2018) derive the 'No Transport, Fixed $\epsilon_{\rm ff}$ ' model by setting $\sigma_{v,z} = \sigma_{v,\rm sf}$. In that case, the contribution is purely driven by the balance between gravitational collapse and star formation driven by supernovae outwards. The model is similar to the model proposed by Ostriker & Shetty (2011).

The 'No Transport, Fixed Q' model, is derived by revisiting their transport equation and looking for solutions where Q is set as a constant. They derive a slightly different relation given by

$$SFR = \frac{4\eta \sqrt{\phi_{mp} \phi_{nt}^3} \phi_Q}{GQ^2} \left(\frac{p_*}{m_*}\right)^{-1} \frac{f_{g,Q}^2}{f_{g,P}} v_c^2 \sigma_{v,z}^2.$$
 (18)

The formulation of different drivers using the same theoretical backing allows for a relatively easy comparison between the observations and different model assumptions.

 Table 3. Parameter values for Krumholz et al. (2018) theoretical model tracks used for Fig. 9.

Parameter	Local dwarf	Local spiral	High z	
$f_{\rm sf}$	0.2	0.5	1.0	
$v_{\rm c} ({\rm km s^{-1}})$	100	220	200	
torb (Myr)	100	200	200	
β	0.5	0.0	0.0	
$f_{g,O} = f_{g,P}$	0.9	0.5	0.7	
ϕ_a	1	1	3	
$SFR_{min} (M_{\odot} yr^{-1})$	-	-	1	
$SFR_{max} (M_{\odot} yr^{-1})$	0.5	50	-	

5.4 Comparison with theoretical model tracks

We now compare our observations to the theoretical models described above. We compare our data to the Krumholz et al. (2018) theoretical model tracks for various galaxy groups; low-*z* dwarfs, low-*z* spirals, and high-*z* galaxies. For each galaxy group, we use the set of parameters suggested by Krumholz et al. (2018), which are shown in Table 3. To account for the thermal and expansion contributions to the velocity dispersion of the H II regions, 15 km s⁻¹ was added in quadrature to the theoretical models.

We find the best agreement between our data and the 'Transport + Feedback' model (Fig. 9). The lower end of the SFR– $\sigma_{v,z}$ relation in the range SFR $\in [10^{-3}, 1] \text{ M}_{\odot} \text{ yr}^{-1}$ is explained by the floor of the 'Transport + Feedback' model tracks, which is driven by star formation feedback processes. Importantly, the slight increase in $\sigma_{v,z}$ can be explained by a change in galaxy properties across the dynamic range of SFR. The upturn in the SFR– $\sigma_{v,z}$ relation at SFR $\geq 1 \text{ M}_{\odot} \text{ yr}^{-1}$ is also consistent with 'Transport + Feedback' model tracks. This is in contrast to the alternative models, which cannot account for the relation across the full dynamic range of SFR.

The 'No Feedback' model is able to model the upturn in the SFR- $\sigma_{v,z}$ relation, but it cannot account for the lower end of the relation. At the lower end of the relation, this model assumes $\sigma_{v,z}$ approaches the thermal and expansion contributions alone. We observed that most of our galaxies lie above the assumed $\sigma_{v,z} > 15 \,\mathrm{km \, s^{-1}}$ contributions from the thermal and expansion broadening. Furthermore, there is a positive correlation of $\sigma_{v,z}$ with SFR even at SFR $\lesssim 10 \,\mathrm{M_{\odot} \, yr^{-1}}$ that the 'No Feedback' model does not appear to account for. Despite the 'No Feedback' model appearing to be a better model, we note that it is difficult to distinguish between the 'No Feedback model' and 'Transport + Feedback' model, as the thermal and expansion broadening contribution is not well known.

The 'No Transport, Fixed $\epsilon_{\rm ff}$ ' model accounts well for the lower end SFR- $\sigma_{v,z}$ relation in our sample. However, it predicts very little evolution in $\sigma_{v,z}$ across galaxy properties for low-*z* galaxies. This is in contrast to the observations that do appear to have an upturn in $\sigma_{v,z}$ for increasing SFR. This suggests that there must be an additional energetic input to the 'No Transport, Fixed $\epsilon_{\rm ff}$ ' to account for increase $\sigma_{v,z}$ across SFR.

The 'No Transport, Fixed Q' model provides an alternative SFR– $\sigma_{v,z}$ relation (SFR $\propto \sigma_{v,z}^2$). The upturn in the theoretical relation qualitatively matches the observed upturn. However, the model tracks are lower than the observed $\sigma_{v,z}$. Similar to the 'No Feedback' model, increasing the thermal and expansion contributions to $\sigma_{v,z}$ would result in better agreement. The 'No Transport, Fixed Q' cannot account for the increased scatter in $\sigma_{v,z}$ for increasing SFR, due to estimating very little variation in $\sigma_{v,z}$ across most of our dynamic range of SFR.



Figure 9. Comparison of the intrinsic vertical velocity dispersion compared to the theoretical model proposed by Krumholz et al. (2018). From left to right, we show the 'Transport + Feedback', 'No Feedback', 'No Transport, Fixed $\epsilon_{\rm ff}$ ', and 'No Transport, Fixed Q' models. The individual tracks use a set of parameters (see Table 3) that represent typical galaxies for each galaxy type. We find that our observations are the most consistent with the 'Transport + Feedback' model.

To distinguish between the 'Transport + Feedback' and 'No Transport, Fixed Q' models, we also compare the theoretical model tracks while varying the circular velocity (see Fig. 10). We see generally good agreement between the 'Transport + Feedback' model tracks and the observed velocity dispersion. The upturn in the velocity dispersion occurs approximately at the expected circular velocity.

To quantify the differences, we calculate the relative residuals between the data and the models. To do this, we used the 'local spiral' tracks for SFR < 10 M_o yr⁻¹ and a model with intermediate parameters between the 'local spiral' and 'high-z' models ($f_{sf} =$ 0.8, $t_{orb} = 200 \text{ M}_{\odot} \text{ yr}^{-1}$, $\beta = 0$, $f_{g,Q} = f_{g,P} = 0.6$, $\phi_a =$ 2) for SFR $\geq 10 \text{ M}_{\odot} \text{ yr}^{-1}$. The relative residuals between the model tracks and data reveal $\Delta \sigma_{v,z}/\sigma_{v,z} = -0.02 \pm 0.32$ for the 'Transport + Feedback' model compared to $\Delta \sigma_{v,z}/\sigma_{v,z} =$ 0.29 \pm 0.42 for the 'No Transport, Fixed Q' model. In particular, the relative residuals for the 'No Tranport, Fixed Q' model increase to $\Delta \sigma_{v,z}/\sigma_{v,z} = 1.16 \pm 0.52$ for SFR > 10 M_☉ yr⁻¹. Thus, suggesting that the 'Transport + Feedback' model provides a better fit to the data than the 'No Transport, Fixed *Q*' model.

For galaxies at SFR $\gtrsim 10 \text{ M}_{\odot} \text{ yr}^{-1}$, we require a transition to values more representative of the high-*z* galaxy model tracks, with higher f_{sf} , $f_{g,Q}$, and $f_{g,P}$ to explain the SFR– $\sigma_{v,z}$ relation. This is not surprising given that those galaxies were selected from the DYNAMO sample. Many of these galaxies exhibit similar properties to those of high-*z* galaxies (Green et al. 2014; Fisher et al. 2017) including increased molecular gas fractions (Fisher et al. 2014).

A similar conclusion was reached by Übler et al. (2019), when comparing the 'Transport + Feedback' model tracks as a function of circular velocity for high-z galaxies. They found ~ 60 per cent of their galaxies could be explained by varying the circular velocity alone.



Figure 10. Comparison of the velocity dispersions for the total galaxy sample to the 'Transport + Feedback' (left) and 'No Transport, Fixed Q' (right) models proposed by Krumholz et al. (2018). The top two panels show the data compared to the model tracks, where the data and model tracks are colour coded by $v_{2,2,\text{tf}}$. For all other input parameters to the model tracks, the solid lines use the 'local spiral'. The dashed lines use intermediate values between the 'local spiral' and 'high-z' models; $f_{\text{sf}} = 0.8$, $t_{\text{orb}} = 200 \text{ M}_{\odot} \text{ yr}^{-1}$, $\beta = 0, f_g, Q = f_g, P = 0.6$, $\phi_a = 2$. See Table 3 for the 'local spiral' and 'high-z' parameters. The bottom two panels show the relative residuals, where $\Delta \sigma_{v,z} = \sigma_{v,z} - \sigma_{v,z,\text{model}}$. We use the models represented by the solid lines for SFR < 10 M_{\odot} yr⁻¹ and the dashed lines for SFR $\geq 10 \text{ M}_{\odot} \text{ yr}^{-1}$. We also show the mean and standard deviation of the relative residuals for each model. Both theoretical models predict an increase in $\sigma_{v,z}$ as a function of SFR, however, 'Transport + Feedback' provides a better fit as a function of circular velocity ($v_{2,2,\text{ tf}}$).

Increasing the molecular gas fraction (f_{sf}) and the gas gravitational contribution at the mid-plane $(f_{g,Q}, f_{g,P})$ also shifts the base $\sigma_{v,z}$ by a few km s⁻¹. As galaxies shift to higher $f_{sf}, f_{g,Q}, f_{g,P}$ as a function of SFR, this provides a mechanism to explain the increase in $\sigma_{v,z}$ seen in the SAMI Galaxy Survey (see Section 3.2.1).

In comparison, the 'No Transport, Fixed *Q*' model predicts an increase in $\sigma_{v,z}$ as a function of SFR at a slower rate than the 'Transport + Feedback' model. Comparing the model tracks when varying the circular velocity and gas properties, we find that the $\sigma_{v,z} \gtrsim 30 \,\mathrm{km \, s^{-1}}$ are not predicted unless assuming a much lower circular velocity ($v_{2.2, \,\mathrm{tf}} \lesssim 50 \,\mathrm{km \, s^{-1}}$) than expected given the stellar masses of the galaxies. Increasing the molecular gas content and gas gravitational contribution at the mid-plane as in the high-*z* galaxies only shifts the model tracks to higher SFR.

The above analysis suggests that the 'Transport + Feedback' model provides a better agreement with the data than those dominated by star formation feedback processes. This does not completely rule out star formation feedback processes as the primary driver, instead it may suggest that the assumed energy momentum due to star formation feedback is too low. The assumed energy source is purely from single supernova, with momentum injection per unit of stars of $\langle p_*/m_* \rangle = 3000 \,\mathrm{km \, s^{-1}}$. However, $\langle p_*/m_* \rangle$ may be significantly higher if other sources are incorporated. For example, Gentry et al. (2017) argue that $\langle p_*/m_* \rangle$ could be up to an order of magnitude higher when incorporating the effects of clustered supernova. As such, further studies will be required to understand the energetic sources of star formation feedback processes to incorporate in these models. As a further caveat to the above analysis, we note that the theoretical models assume that we are observing the star-forming molecular gas, rather than the ionized gas. The full set of differences between the kinematics of the molecular star-forming gas compared to the ionized gas is not complete. For example, there is evidence that ionized gas may have systematically lower rotation and higher velocity dispersions compared to the molecular gas (Levy et al. 2018). However, there is limited research into these differences at this time, as such we make the assumption that these differences are minimal. Further research into the differences in molecular gas and ionized gas kinematics will be required.

5.5 Comparing the correlation analysis to the theoretical models

The above theoretical models (equations 16 and 18) suggest that SFR $\propto v_c^2$, all else being set equal. Thus, we should expect a strong inverse relationship between $\sigma_{v,z}$ and v_c . In Fig. 7, we showed that there is a negative correlation between velocity dispersion and rotational velocity after accounting for the stellar mass contribution. We are forced to control for the stellar mass using the Tully–Fisher relation as both $\sigma_{v,z}$ and v_c increase for increasing stellar mass.

As such, the rotational velocity is a significant factor in prescribing the intrinsic turbulence within the galaxy. This is consistent with the theoretical models of Krumholz et al. (2018). However, the relationship between the turbulence and rotational velocity does not distinguish between star formation feedback or gravitational driven mechanisms of turbulence. The proposed models also suggest a dependence of the SFR– $\sigma_{v,z}$ relation on the mid-plane gas fraction $(f_{g,P})$, the mid-plane gas contribution to the toomre-Q parameter $(f_{g,Q})$, and on the molecular to neutral gas fraction (f_{sf}) . Krumholz et al. (2018) also showed that galaxy turbulence driven solely by star formation feedback has the relation SFR $\propto \sigma_v f_{g,Q}^2 / f_{g,P}$, whereas solely driven by gravitational mechanisms has SFR $\propto \sigma_v f_{g,Q}^2$.

The contribution of the gas content to the velocity dispersion is difficult to determine in our sample. We have measurements of the integrated H_I mass for 95 galaxies in our sample from the SAMI Galaxy Survey. We showed a slight negative but still consistent with zero correlation between the total H_I gas fraction (f_g) and $\sigma_{v,z}$ in Section 3.2.1.

A negative correlation between integrated HI mass and $\sigma_{v,z}$ could be due to the expected negative correlation expected between $\sigma_{v,z}$ and $f_{g,Q}$ in the 'Transport + Feedback' model. However, it could also be a result of increasing molecular gas fraction (f_{sf}) for increasing SFR and M_{*} that are also positively correlated with $\sigma_{v,z}$. We also note that the integrated HI measurements are not the ideal measurement as we cannot determine the mid-plane HI gas content within each galaxy. To accurately determine the relation between $\sigma_{v,z}$ and the gas content of the galaxy, we expect that resolved measurements of the HI and H₂ masses are required. In that way, we would be able to more precisely determine the midplane gravitational contribution of the galaxy gas content. We note that recent work by Sun et al. (2020) has begun to shed light on the mid-plane gas contributions to the observed turbulence, although further studies will be required.

6 CONCLUSIONS

We studied the intrinsic kinematic properties of the ionized gas in 383 low-*z* star-forming galaxies. A total of 342 galaxies were obtained from the SAMI Galaxy Survey DR2 plus another 41 were from the DYNAMO survey. The total galaxy sample spans a wide range of galaxy properties with SFR $\in [10^{-3}, 10^2] \text{ M}_{\odot}$ yr⁻¹. The intrinsic gas kinematics were estimated using BLOBBY3D. BLOBBY3D is a flexible galaxy modelling approach that assumes that the galaxy is regularly rotating with spatially clumpy ionized gas distributions. In order to mitigate the effects of beam smearing and instrumental broadening, a convolution by the PSF and LSF on the underlying model is performed prior to calculating the likelihood function. We also performed a minor inclination correction for the sample from the SAMI Galaxy Survey to estimate the intrinsic vertical velocity dispersion ($\sigma_{v,z}$) as described in Section 3.2.1.

The sample of galaxies from the SAMI Galaxy Survey is a representation of typical galaxies at $z \leq 0.1$. As such, we only used that galaxy sample to determine the typical gas kinematics in galaxies at $z \leq 0.1$. We find the following:

(i) Low velocity dispersions of $\sigma_{v,z} \in [14.1, 22.1] \text{ km s}^{-1}$ for the 68 per cent shortest credible interval. This is ~10 km s⁻¹ lower than previous studies of the SAMI Galaxy Survey. The difference in results is likely driven by our beam smearing correction technique using BLOBBY3D, compared to the heuristic approaches applied by Zhou et al. (2017) and Johnson et al. (2018). We also find little evidence for a significant population of galaxies with $\sigma_{v,z} \gtrsim$ 50 km s⁻¹ as found by Yu et al. (2019) in a sample of galaxies of similar galaxy properties from the MaNGA Survey. In contrast, our velocity dispersions are approximately consistent with other studies of nearby galaxies (Epinat et al. 2008; Moiseev et al. 2015). (ii) There is a significant positive correlation between $\sigma_{v,z}$ and SFR measures. The greatest correlation was with Σ_{SFR} . Although the correlation is significant, the average $\sigma_{v,z}$ only increased by $\sim 6 \text{ km s}^{-1}$ for a dynamic range of SFR $\in [10^{-3}, 10] \text{ M}_{\odot} \text{ yr}^{-1}$.

(iii) We also find positive correlations of $\sigma_{v,z}$ with integrated stellar and H I gas mass as well as absolute rotational velocity.

(iv) After controlling for stellar mass, there is a negative correlation between $\sigma_{v,z}$ and rotational velocity. This is consistent with theoretical models proposed by Krumholz et al. (2018) for both star formation feedback processes and gravitational driving mechanisms of turbulence.

(v) We find a weak, but still consistent with zero, negative trend between $\sigma_{v,z}$ and the integrated HI gas fraction. Theoretical models have suggested that there should be a relation between the gravitational contributions of the gas at the mid-plane and $\sigma_{v,z}$. We suspect that the signal between gas fraction and $\sigma_{v,z}$ is lost when using the integrated HI mass. Accurately determining the gravitational contributions of both HI and H₂ at the mid-plane is likely required to observe the proposed relations.

The combined SAMI Galaxy Survey and DYNAMO data sets span a wide range of SFR, allowing for improved comparisons to the theoretical models proposed by Krumholz et al. (2018). The SFR– $\sigma_{v,z}$ relation for our sample of galaxies is the most consistent with the 'Transport + Feedback' model proposed by Krumholz et al. (2018). We find that the SFR– $\sigma_{v,z}$ relation can be approximately explained by a transition of increasing circular velocity and molecular gas at higher SFR.

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MNRAS 495, 2265-2284 (2020)

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SUPPORTING INFORMATION

Supplementary data are available at *MNRAS* online.

GasTurbulenceDriversAppendix

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