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Galaxy And Mass Assembly (GAMA): Bivariate functions of $H\alpha$ star forming galaxies

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ABSTRACT

We present bivariate luminosity and stellar mass functions of $H\alpha$ star forming galaxies drawn from the Galaxy And Mass Assembly (GAMA) survey. While optically deep spectroscopic observations of GAMA over a wide sky area enable the detection of a large number of $0.001 < SFR_{H\alpha} (M_{\odot} \text{ yr}^{-1}) < 100$ galaxies, the requirement for an $H\alpha$ detection in targets selected from an r -band magnitude limited survey leads to an incompleteness due to missing optically faint star forming galaxies. Using $z < 0.1$ bivariate distributions as a reference we model the higher- z distributions, thereby approximating a correction for the missing optically faint star forming galaxies to the local SFR and \mathcal{M} densities. Furthermore, we obtain the r -band LFs and stellar mass functions of $H\alpha$ star forming galaxies from the bivariate LFs. As our sample is selected on the basis of detected $H\alpha$ emission, a direct tracer of on-going star formation, this sample represents a true star forming galaxy sample, and is drawn from both photometrically classified blue and red sub-populations, though mostly from the blue population. On average 20–30% of red galaxies at all stellar masses are star forming, implying that these galaxies may be dusty star forming systems.

Key words: surveys – galaxies: luminosity functions – galaxies: evolution – galaxies

1 INTRODUCTION

The observed univariate luminosity function (LF) is one of the fundamental measures of galaxy properties. It is usually one of the first results to be measured from galaxy surveys (e.g. Davis & Huchra 1982; Loveday et al. 1992; Lin et al. 1996; Norberg et al. 2002; Blanton et al. 2003b; Loveday et al. 2012). The importance of the LF, defined as the co-moving source density with luminosity (or magnitude) $L + \Delta L$, extends to all areas of astronomy. In an observational context, it is used to quantify the mean space density of galaxies per unit luminosity and the evolution of statistical

properties of a galaxy sample across cosmic time (e.g. Ly et al. 2007; Dale et al. 2010; Westra et al. 2010; Gunawardhana et al. 2013; Drake et al. 2013; Sobral et al. 2013, 2014). In theoretical modelling, the LF is a key ingredient needed to constrain the dark matter halo formation (e.g. Croton et al. 2006; Bower et al. 2008). In an era of multi-wavelength legacy surveys with intrinsically complicated multi-band selections (e.g. Driver et al. 2011, Galaxy And Mass Assembly survey), understanding the effects of selection and systematic biases on the shape of the LF is imperative in obtaining reliable LF measurements.

Simply due to the existence of detection limits, no single survey can directly detect all sources to provide a complete and unbiased galaxy sample. The detection probability of an object is a function of a number of parameters, both external (e.g. survey selection,

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area and depth, star–galaxy separation, observing conditions and redshift, spectroscopic and target completeness) and intrinsic to the object (e.g. surface brightness, size and colour). As luminosity is strongly correlated with both sets of factors, the least luminous objects in any magnitude–limited survey have the poorest detection probabilities (Geller et al. 2012), thus occupying a relatively small volume. The low luminosity galaxies, although they do not dominate the luminosity budget of the universe, greatly outnumber the luminous giants. As other studies have emphasised (Petrosian 1998), measuring the evolution of the slope of the faint end of the LF is a challenge. This arises because of the preferential bias against faint galaxies due to surface brightness limits (Sprayberry et al. 1996; Dalcanton et al. 1997; Geller et al. 2012), galaxy morphologies (Marzke et al. 1998; Tempel et al. 2011), spectral types (Folkes et al. 1999; Madgwick et al. 2002), environment (Xia et al. 2006; Tempel et al. 2009; Zandivarez & Martínez 2011) and colour (Blanton et al. 2001) as well as external issues (Driver et al. 2005; Loveday et al. 2012).

Any galaxy sample selected based on a parameter other than the primary survey selection criteria is biased as a result of the dual sample and survey selection. Gunawardhana et al. (2013) and Westra et al. (2010) present the $H\alpha$ univariate LFs and determine the evolution of $H\alpha$ star formation rate density (SFRD) in the local universe using $H\alpha$ star forming (SF) galaxy samples drawn respectively from the r -band magnitude–limited Galaxy And Mass Assembly (GAMA) and Smithsonian Hectospec Lensing surveys. These studies show that their lowest redshift ($z < 0.1$) samples are in fact the most complete and span the largest range in both intrinsic $H\alpha$ luminosity ($L_{H\alpha}$) and r -band absolute magnitude (M_r), e.g. the GAMA $z < 0.1$ sample probes $30.5 \lesssim \log L_{H\alpha} (W) \lesssim 36$, $-24 \lesssim M_r \lesssim -10$ (Gunawardhana et al. 2013) and $7 \lesssim \log \mathcal{M}/M_\odot < 12$, where \mathcal{M} is the galaxy stellar mass (Baldry et al. 2012). With increasing redshift, however, the sample completeness drops in the sense that a fraction of optically faint star forming galaxies are missing from the higher- z subsamples and this fraction increases with increasing redshift. As a consequence the final SFRDs based on bivariate selected samples are underestimated, manifesting as an apparent lack of evolution with redshift in contrast to current observations (Hopkins & Beacom 2006; Pérez-González et al. 2008; Karim et al. 2011; Sobral et al. 2014). In comparison, the LFs based on narrowband surveys do not suffer from the same bias as their targets are selected using the quantity they aim to measure (see Gunawardhana et al. 2013, for a discussion on the advantages and disadvantages of broadband and narrowband surveys). In case of a bivariate selected sample, as is the case in this study, to recover the missing contribution from optically faint star forming galaxies requires studying how the selection biases influence the LF.

The bivariate LF (Phillipps & Disney 1986) provides a powerful method of studying the luminosity density in different epochs inclusive of selection biases (e.g. bivariate brightness distributions, bivariate luminosity and size distribution). There is a rich collection of literature on using bivariate LFs to explore the space density of galaxies as a function of both survey selection wavelength and surface brightness limits (e.g. Driver 1999; Cross et al. 2001; Blanton et al. 2001; Driver et al. 2005), galaxy size (e.g. Sodre & Lahav 1993; de Jong & Lacey 2000; de Jong et al. 2004; Cameron & Driver 2007), radio luminosity (e.g. Sadler et al. 1989; Ledlow & Owen 1996; Mauch & Sadler 2007), Sérsic index, stellar mass and spectral type (e.g. Ball et al. 2006a), colour (e.g. Baldry et al. 2004) and in pairs of various galaxy properties (e.g. Blanton et al. 2003a; Driver et al. 2006) as well as bivariate ultraviolet/infrared LFs (e.g. Saunders et al. 1990; Takeuchi et al. 2012).

In this followup paper to Gunawardhana et al. (2013), hereafter paper I, we explore the GAMA bivariate $L_{H\alpha}$ – M_r and $L_{H\alpha}$ – \mathcal{M} functions. Paper I presents the local star formation history as traced by $H\alpha$ emitters contained within photometrically selected GAMA galaxies. The GAMA $H\alpha$ LFs probe a wider range in luminosity than other results to-date and demonstrate a Gaussian-like drop in number density (Φ) at high luminosities, rather than the exponential drop characteristic of Schechter (1976) function. In paper I we conclude that a Saunders et al. (1990) functional form, widely used to characterise radio and infrared LFs, is now also required to give a better description of $H\alpha$ LFs. Despite the relatively large range in luminosity probed by the GAMA $H\alpha$ LFs up to $z < 0.34$, the intrinsic SFR densities based on these LFs show a distinct lack of evolution in SFR density with increasing redshift. This apparent lack of evolution in SFRD is primarily caused by the sample selection. For GAMA, we find that the r -band apparent magnitude limit of the survey, along with the subsequent requirement for $H\alpha$ detection, leads to an incompleteness due to missing bright $H\alpha$ sources that are fainter than the r -band selection limit. While local estimates of SFR density measurements from a range of SFR-sensitive wavelengths (e.g. $H\alpha$, [O II], [O III], $H\beta$) based on narrowband surveys and slitless spectroscopy data show an evolution with redshift (e.g. Jones & Bland-Hawthorn 2001; Shioya et al. 2008; Sobral et al. 2013), those based on broadband surveys are almost always underestimated (e.g. paper I; Westra et al. 2010) as a consequence of the bivariate sample selection. The primary aims of this investigation are to: (a) model the low redshift bivariate function to use as a reference to account for the missing optically faint star forming galaxies at higher- z ($z < 0.34$), (b) measure the (moderate) redshift evolution of stellar mass and SFR densities, (c) characterise the univariate M_r LFs and stellar mass functions (SMF) of $H\alpha$ SF galaxies and compare the results with LFs and SMFs of photometrically classified blue galaxies, and (d) explore the characteristics of photometrically classified blue and red star forming sub-populations in GAMA.

The layout of this paper is as follows. We briefly describe the GAMA survey, our sample selection criteria and selection biases in § 2. In § 3, we describe the derivation of the bivariate functions, taking into account different survey selection criteria. The resulting bivariate functions are presented in § 4 and § 6. These sections also include the univariate functions obtained from integrating the bivariate functions over $L_{H\alpha}$ axis. Finally, in § 5 we detail of the functional forms used to fit the bivariate functions, and in § 5.2 and § 6.3, we infer SFR and \mathcal{M} densities by integrating the best-fitting functional forms to the bivariate functions. Our results are discussed in § 7 and we conclude in § 8.

The assumed cosmological parameters are $H_0 = 70 \text{ km s}^{-1} \text{ Mpc}^{-1}$, $\Omega_M = 0.3$, and $\Omega_\Lambda = 0.7$. All magnitudes are presented in the AB system. A Chabrier (2003) IMF, commonly used to derive stellar masses in the literature, is used to derive the stellar mass measurements used in this study (Taylor et al. 2011) and the same IMF used in paper I (i.e. Baldry & Glazebrook 2003, IMF) is used to estimate SFRs. As these two IMF forms are sufficiently similar, no significant systematic effect introduced by adopting them. To avoid confusion we state in the figure caption which IMF is used to obtain the results shown.

2 THE GAMA SURVEY AND BIVARIATE SAMPLE SELECTION

Our study utilises data from the GAMA¹ survey (Baldry et al. 2010; Robotham et al. 2010; Driver et al. 2011). GAMA is a spectroscopic survey undertaken at the Anglo Australian Telescope (AAT) with 2-degree Field (2dF) fibre feed and the AAOmega multi-object spectrograph. AAOmega provides a resolution of 3.2 Å full width at half-maximum with complete spectral coverage from 3700 to 8900 Å (Sharp et al. 2006; Driver et al. 2011; Hopkins et al. 2013). GAMA-I covers three equatorial fields at approximately 09hr, 12hr and 15hr, hereafter G09, G12 and G15, of 48 deg² each. G09 and G15 are limited to a depth of $r_{AB} < 19.4$ while G12 extends to $r_{AB} < 19.8$.

For the current analysis, we use GAMA-I spectroscopic data consisting of GAMA, SDSS, 2-degree Field Galaxy Redshift Survey (Colless et al. 2001, 2dFGRS) and Millennium Galaxy Catalogue (Driver et al. 2005, MGC) sources. A detailed description of the data, the sample selection and the measurement of physical properties of galaxies such as SFRs is given in § 2 of paper I. Briefly, the emission line measurements used for this investigation are measured from each flux calibrated spectrum, assuming a single Gaussian profile and a common redshift and line width within an adjacent set of lines (e.g. H α , the [N II] $\lambda\lambda$ 6548, 6583 and [S II] doublets), and simultaneously fitting the continuum local to the set of lines. This method of measuring fluxes does not take into account the effects of the underlying stellar absorption on Balmer line fluxes. To correct for this effect, we apply a constant correction to H α and H β fluxes following the prescription of Hopkins et al. (2003). A comprehensive discussion of GAMA spectroscopic measurement process is presented in Hopkins et al. (2013), and further analyses on the effects of the assumption of a constant stellar absorption correction are presented in paper I and in Gunawardhana et al. (2011).

The derivation of stellar absorption, aperture and dust obscuration corrected H α luminosities ($L_{H\alpha}$) is described in § 3 of paper I. The stellar masses (\mathcal{M}) and absolute magnitudes k -corrected to $z = 0$ (M_r) have been derived using the stellar template spectrum that best fits u, g, i, r, z GAMA photometry (StellarMassesv08, Taylor et al. 2011). We have not attempted to correct the rest frame r -band continuum flux for H α emission contamination as the contamination is $< 2\%$ for over 98% of the lowest redshift data (H α spectral line is redshifted out of the r -band filter at $z \sim 0.05$), and does not change the results we present in this analysis. Additionally, we compute absolute magnitudes based on SDSS Petrosian r -band magnitudes at $z = 0.1$, hereafter $M_r^{0.1}$ (KCORRECT V4.2, Blanton & Roweis 2007) to compare our results with previous GAMA studies (e.g. Loveday et al. 2012; Baldry et al. 2012).

We use the same sample of galaxies as in paper I. Briefly, our sample includes all emission-line galaxies with H α fluxes ($F_{H\alpha}$) greater than a detection limit of 25×10^{-20} W/m², and H α emission signal-to-noise greater than 3 that are classified as star forming based on the prescription of Kewley & Dopita (2002). The emission line measurements for galaxies in the three GAMA regions that have been observed previously in earlier spectroscopic campaigns (e.g. SDSS, 2dFGRS, 6dFGRS) are either taken from their respective survey databases or accounted for through spectroscopic incompleteness corrections if the respective spectra are not flux calibrated (e.g. 2dFGRS spectra). The sample properties and trends with redshift are explored in § 2 and § 3 of paper I. To

summarise, the sample used for the calculation of $f_{V,H\alpha}$ includes the following galaxies:

1. GAMA observations with observed H α emission above the detection limit of 25×10^{-20} W/m². AGNs are removed from the sample using BPT diagnostics and the Kewley & Dopita (2002) prescription. The two line classifications (paper I; Brough et al. 2011) are used to recognise AGNs if one of the four spectral line measurements required for a BPT diagnostics is unavailable. A small subset of GAMA galaxies where none of these methods can be employed is included in the sample, as these are more likely to be star forming than AGNs (Cid Fernandes et al. 2011).
2. SDSS observations with H α emission and continuum signal-to-noise greater than 3. The AGNs are removed as described above using SDSS line measurements available from MPA-JHU database².

Figures 1 and 2 demonstrate the limitations of a bivariate selected sample. Figure 1 shows the M_r distributions of all GAMA galaxies in four different redshift ranges compared to the distributions of galaxies with reliable H α detections (i.e. $F_{H\alpha} > 25 \times 10^{-20}$ W/m²) and galaxies with $F_{H\alpha} > 1 \times 10^{-18}$ W/m² (i.e. the flux limit used in the V_{\max} calculations; § 4 of paper I). The distributions of blue and red sub-populations, as defined in Eq. 4 of this paper and in Loveday et al. (2012), of galaxies within each distribution are shown in the insets. The distributions of H α detected galaxies at each redshift range comprise both blue and red galaxies, though blue galaxies dominate the distributions. Also, the $z < 0.1$ and $0.1 < z < 0.15$ distributions of the H α detected galaxies show a clear blue-red bimodal distribution, while the lack of such a trend in the higher redshift distributions is likely partly due to the difficulty in reliably measuring the H α feature in low signal-to-noise weak-line systems at higher redshifts, and partly due to the smaller luminosity range probed at higher redshifts. Even though the distribution of H α detected sample is bimodal, the sample, after imposing an H α flux limit for the estimation of V_{\max} (paper I), is biased against red weak-line galaxies at all redshifts, which are likely to be dusty star forming galaxies with low signal-to-noise spectra (§ 6.1).

Figure 2 further demonstrates how the bivariate selection acts to limit the range of $L_{H\alpha}$ and M_r probed by the LFs with increasing redshift. The $z < 0.1$ sample probes the largest range in both M_r and $L_{H\alpha}$. The dual H α - M_r selection prevents the optically faint star forming galaxies from entering the sample at all redshifts, with its effects becoming more significant with increasing redshift. In order to assess the extent of this effect, we now consider the bivariate LFs.

3 CONSTRUCTING BIVARIATE FUNCTIONS

We use three different LF estimators (the classical $1/V_{\max}$ method, the density corrected $1/V_{\max}$ method and the bivariate step-wise maximum likelihood) to derive the bivariate LFs presented in § 4. The formulation of each of the three methods is described below.

3.1 The “classical” method

The $1/V_{\max}$ technique (Schmidt 1968), also referred to as the “classical method”, is widely used to estimate the co-moving space density as a function of luminosity.

¹ <http://www.gama-survey.org/>

² http://www.mpa-garching.mpg.de/SDSS/DR7/raw_data.html

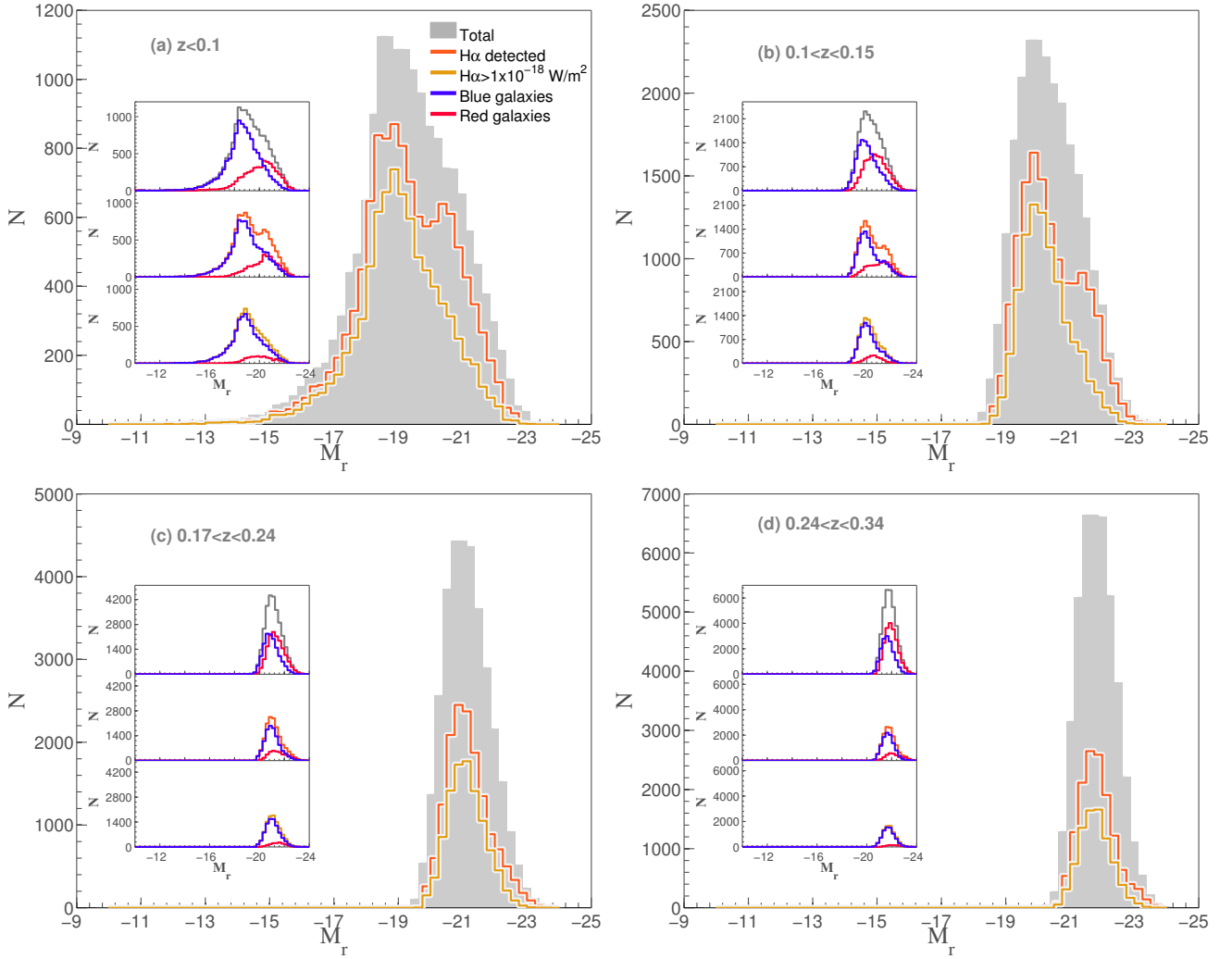


Figure 1. The M_r distribution of GAMA galaxies split in to four redshift bins. Each panel shows three distributions: all galaxies (grey histogram), those with observed $H\alpha$ flux greater than the detection limit of $25 \times 10^{-20} \text{ W/m}^2$ (orange histogram), and those with $H\alpha$ fluxes $> 1 \times 10^{-18} \text{ W/m}^2$, i.e. the flux limit (yellow histogram). Each of the three histograms are further divided to indicate the distributions of galaxies classified as blue and red based on Eq. 4. These are shown within the three insets in each main panel. The three insets top-to-bottom show the distributions of blue and red sub-populations corresponding to all galaxies (top), galaxies with $H\alpha$ fluxes $> 25 \times 10^{-20} \text{ W/m}^2$ (middle) and galaxies with $H\alpha$ fluxes $> 1 \times 10^{-18} \text{ W/m}^2$ (bottom).

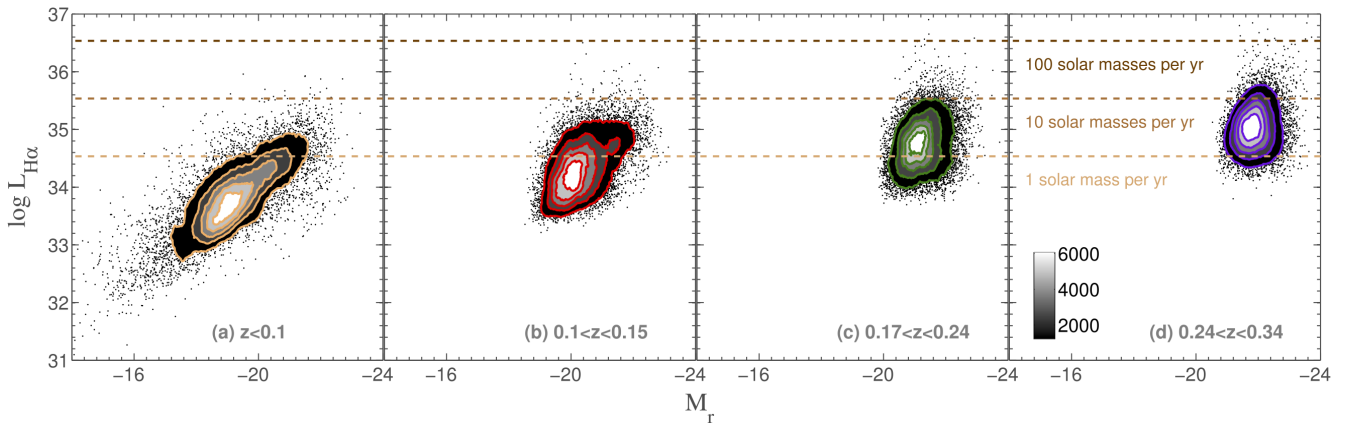


Figure 2. The bivariate $\log L_{H\alpha}$ (W) and M_r distributions split in to four redshift bins. The dual r -band apparent magnitude and $H\alpha$ flux selection of our sample leads to an incompleteness in optically faint star forming galaxies. The $z < 0.1$ sample probes the largest range in $H\alpha$ luminosity and M_r , therefore it is the most complete. The range probed by the higher redshift samples progressively drops with redshift.

The formulation of the $1/V_{\max}$ method inclusive of incompleteness corrections for univariate LFs is described in §4 of paper I. The bivariate function derived using this definition is

$$\Phi[\log L_{H\alpha}, M_r] \times \Delta \log L \Delta M_r = \sum_i \frac{1}{V_{i,\max}}. \quad (1)$$

In this equation, $V_{i,\max}$ represents the maximum volume out to which the i^{th} object would be visible and still be part of the survey, $\Delta \log L$ and ΔM_r define the luminosity and magnitude bin widths, respectively.

The GAMA-I sample used in this study is subjected to two different r -band magnitude limits ($r < 19.4$ for G09, 15 and $r < 19.8$ for G12; Driver et al. 2011) and an emission-line selection. Given these constraints, the definition of $V_{i,\max}$ is

$$V_{i,\max} = \min[(V_{i,\max,H\alpha}), (V_{i,\max,r}), (V_{i,z_{lim}})] \times c_i, \quad (2)$$

where $V_{i,\max}$ is the minimum volume that a galaxy i would have given the maximum volumes for that galaxy computed using an $H\alpha$ flux limit of $1 \times 10^{-18} \text{ W m}^{-2}$ ($V_{i,\max,H\alpha}$) and the magnitude limit of the GAMA survey ($V_{i,\max,r}$), and the volume ($V_{i,z_{lim}}$) of the redshift slice (z_{lim}) that i^{th} galaxy resides in. The completeness correction, c_i , takes into account both the imaging and spectroscopic incompletenesses. The estimation of c_i is described in detail in paper I and in Loveday et al. (2012).

The classical method has the advantages of simplicity and it gives simultaneously both the shape and the absolute normalisation of the LF, however, it can be susceptible to cosmic (sample) variance (Efstathiou et al. 1988; Willmer 1997; Baldry et al. 2012) when the survey volume is small. For this reason, large datasets covering a substantial portion of the sky are generally required to avoid the shape of the LF being distorted due to large scale structure.

There are more sophisticated methods of estimating LFs to account for these disadvantages. Two such methods are the density corrected $1/V_{\max}$ corrections (Baldry et al. 2006; Mahtessian 2011; Cole 2011; Baldry et al. 2012) and the stepwise maximum likelihood method (SWML; Efstathiou et al. 1988), a variant of the method proposed by Sandage et al. (1979).

Finally the univariate LF is obtained by integrating Eq. 1 over one of the variables.

3.2 Density corrected maximum volume corrections

Baldry et al. (2006) (but also Mahtessian 2011; Baldry et al. 2012) describe a modification to the $1/V_{\max}$ technique that takes into account the radial variation in the large-scale structure. They define a maximum volume weighted by density, V'_{\max} ,

$$V'_{i,\max} = \frac{\rho_{ddp}(z_{min}; z_{max,i})}{\rho_{ddp,mean}(z_1; z_2)} \times V_{i,\max} = f_V \times V_{i,\max}, \quad (3)$$

where $\rho_{ddp}(z_1; z_2)$ is the number density of a density defining population (DDP) between redshifts z_1 and z_2 . DDP is a volume limited sample and z_1, z_2 are the minimum and maximum redshifts of that volume limited sample (Baldry et al. 2006, 2012). $\rho_{ddp}(z_1; z_2)$ is estimated separately for each GAMA field, and the average between the three fields is taken to be $\rho_{ddp,mean}$.

As we investigate the evolution of the bivariate $L_{H\alpha}$ - M_r function over a moderate range in redshift, the density weights (f_v) for $V_{\max,r\text{-band}}$ and $V_{\max,H\alpha}$ are estimated separately using several overlapping volume limited samples to improve the accuracy of $V'_{\max,i}$ (Mahtessian 2011). The GAMA sample selection criteria detailed in Baldry et al. (2012), not restricted to their redshift range, is adopted to estimate density weights for $V_{\max,r\text{-band}}$.

We use $f_{v,r}$ to denote the density weights estimated using this sample, which has a univariate r -band apparent magnitude selection. Figure 3a shows the volume limited sample definitions used in the calculation of $f_{v,r}$ for the $z < 0.1$ sample. The variation in $f_{v,r}$ is quantified separately for the G09 and G15 (Figure 3b), and G12 (Figure 3c) fields because of their different magnitude limits. In the regions where two volume limited samples are allowed to overlap, shown as hatched and shaded regions in Figure 3, the $f_{v,r}$ corrections are combined such that the transition from one volume limited sample to the other is smooth. The blue line in Figure 3b and 3c show the final $f_{v,r}$ versus redshift relation used in the calculation of $V'_{\max,r}$ offset by -0.2 in log space for legibility. Also shown for comparison is the $f_{v,r}$ relation from Baldry et al. (2012) for $z < 0.06$ GAMA galaxies (black line). The negligible difference between the blue and black lines is a result of the different r -band magnitude types used as inputs to KCORRECT V4.2 (Blanton & Roweis 2007).

As the galaxy sample used for this study has a dual r -band magnitude and $H\alpha$ flux selection, we also explored the impact of estimating the density corrections using only the $H\alpha$ detections in a r -band selected sample. We use $f_{v,H\alpha}$ to denote this correction. This serves, in principle, as a different correction, $f_{V,H\alpha}$, for $V_{\max,H\alpha}$, although it should be similar in practice. To estimate $f_{V,H\alpha}$, we add 2dFGRS observations³ with η , a measure of the average absorption/emission line strength of a galaxy that strongly correlates with $H\alpha$ equivalent width, greater than -1.2 (Madgwick et al. 2002) to the SF galaxy sample used for this study (§2). The AGNs are removed as described above using the 2dFGRS spectral line catalogue (Lewis et al. 2002).

The distribution of r -band absolute magnitudes in redshift for the selected GAMA $H\alpha$ emission line galaxies with spectra originating from GAMA, SDSS and 2dFGRS surveys is shown in Figure 4a, with the GAMA survey providing the deepest spectroscopic observations followed by 2dFGRS and SDSS surveys. The same volume limited sample definitions introduced in Figure 3 are used to calculate $f_{V,H\alpha}$, but from a r -band magnitude versus redshift distribution comprising only $H\alpha$ SFR galaxies (Figure 4b). The black points in Figure 4b indicate galaxies with spectra originating from a survey other than GAMA, 2dFGRS or SDSS for which we currently do not have spectral line information, and are not included in the sample used to determine $f_{V,H\alpha}$. The spectroscopic incompleteness arising from the exclusion of these objects leads to a small discrepancy between $f_{v,r}$ and $f_{v,H\alpha}$ as demonstrated in Figure 4c. Even though this discrepancy is largest at the lowest redshifts, as might be expected due to the relatively low number of galaxies sampled by the lowest redshift volume-limited sample in comparison to other volume-limited samples, it has a negligible effect on the density corrected bivariate functions.

An alternative approach that does not require binning was developed by Sandage et al. (1979) (STY). The STY maximum likelihood LF estimator, although not biased by the presence of the large scale structure, requires the assumption of a parametric form for the LF. A “nonparametric” variant of STY method called the stepwise maximum likelihood (SWML) estimator was introduced by Efstathiou et al. (1988), mainly to overcome the inconvenience of not being able to adequately establish whether the chosen parameterisation represents a good fit to the data (Efstathiou et al. 1988; Willmer 1997). While this technique is insensitive to density fluctuations (Sandage et al. 1979; Efstathiou et al. 1988; Willmer 1997), the luminosity bins in SWML methods are highly correlated such that any issue that occurs in a given bin may affect the

³ www2.aao.gov.au/2dfgrs/

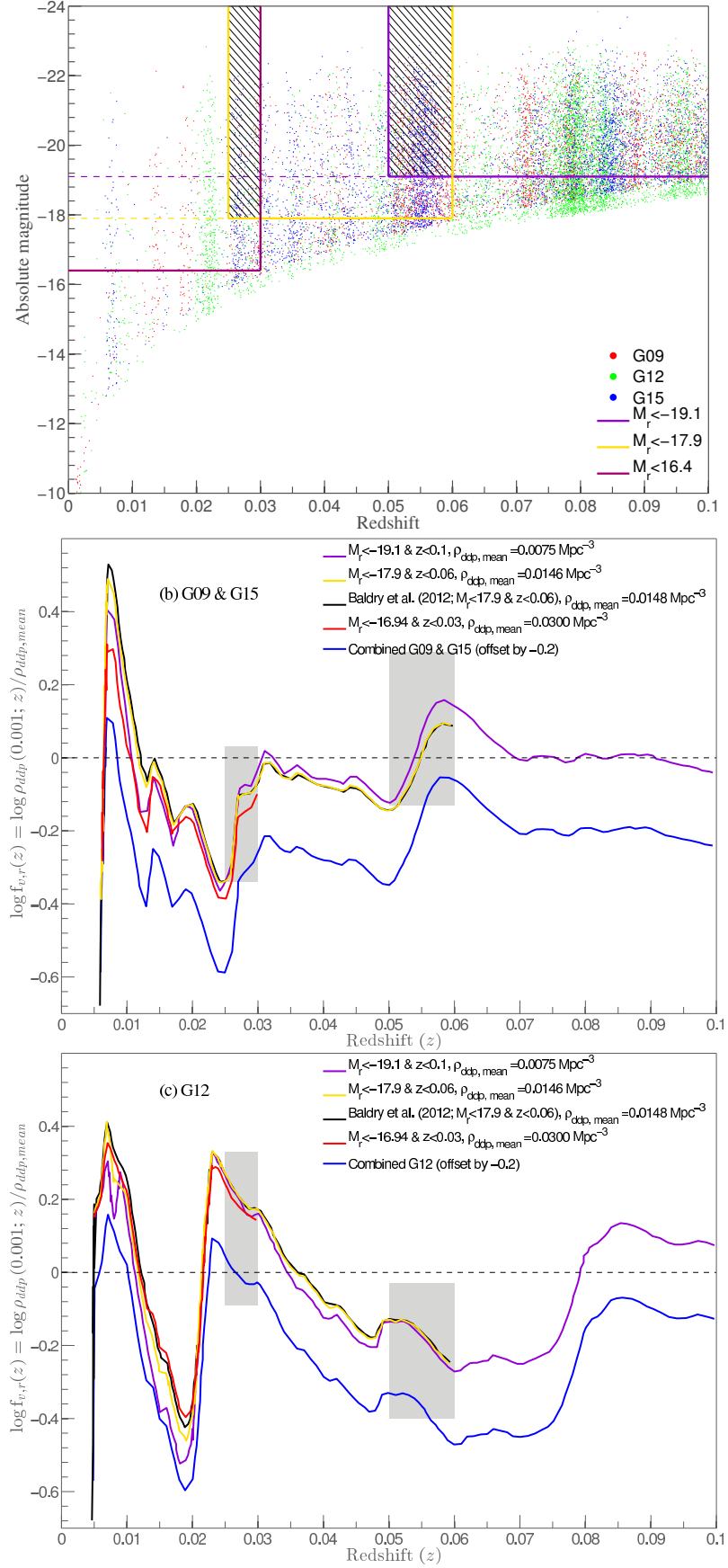


Figure 3. (a) The distribution of absolute r -band magnitudes in redshift for *all* $z < 0.1$ galaxies colour coded by their GAMA region identifier. Using this distribution we define three overlapping volume limited samples, which are then used in the derivation of $\log f_{v,r}$ as a function of redshift for each volume limited sample (b and c). As the three GAMA fields have two different magnitude limits, $f_{v,r}$ is derived separately for G09 and G15 ($r < 19.4$), and G12 ($r < 19.8$). The blue line, offset from the rest for legibility, indicates the variation in final $f_{v,r}$ used for the analysis. This line traces the $f_{v,r}$ versus redshift relation estimated from the lowest redshift volume sample (i.e. red line) until $z \sim 0.03$ (i.e. first shaded region corresponding to the hatched region in the top panel) before smoothly transitioning to the relation derived from the second volume-limited sample (yellow line) and so on. The $f_{v,r}$ versus redshift relations derived in Baldry et al. (2012) for $z < 0.06$ galaxies are shown for comparison.

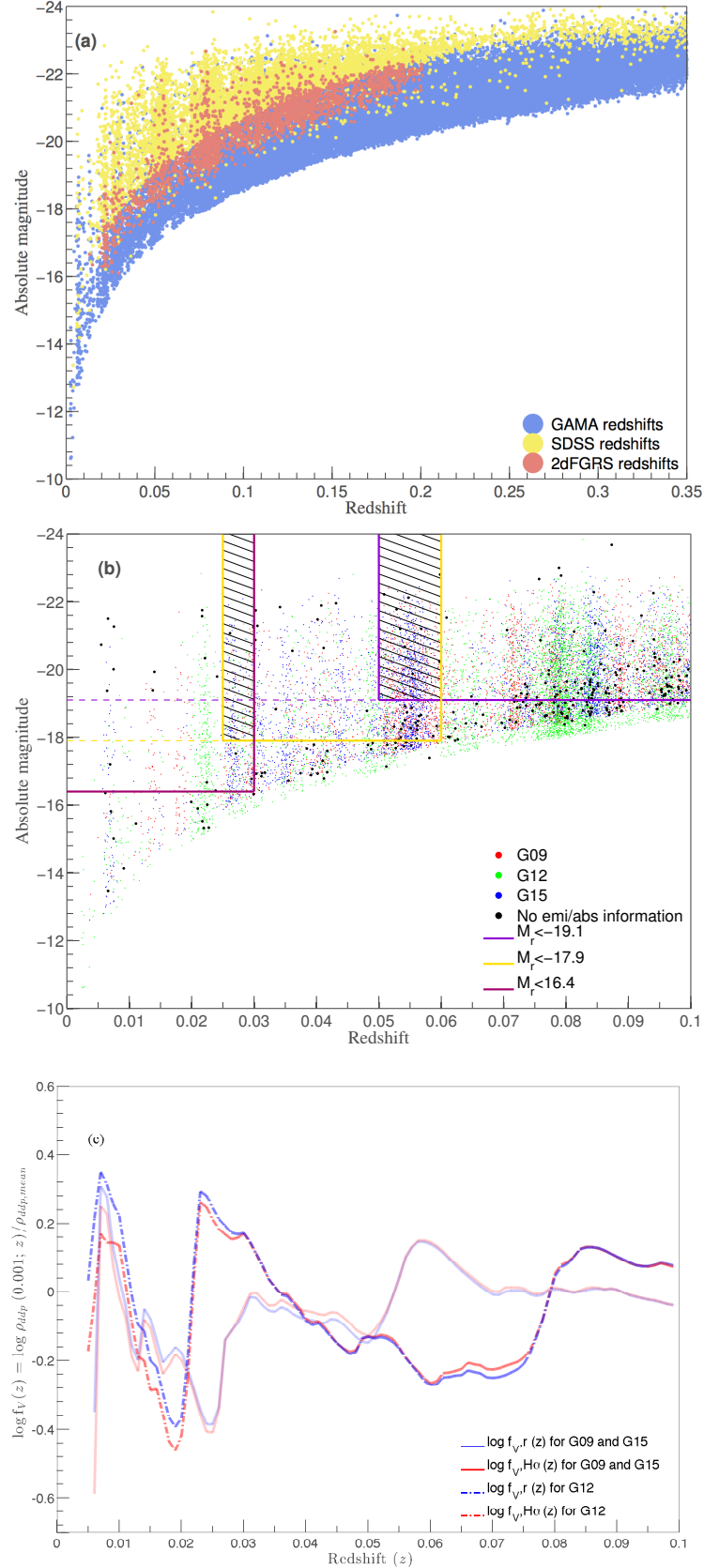


Figure 4. (a) The distribution of absolute r -band magnitudes of $H\alpha$ SFR galaxies in redshift, colour coded by the survey from which the spectra are taken. (b) Same as Figure 3, but for SF galaxies. The black points indicate the absolute magnitudes of galaxies for which we do not have any spectra. (c) The $f_{v,H\alpha}$ (red) and $f_{v,r}$ (blue) versus redshift relations for G09 and G15 (solid line), and G12 (dashed line). The lack of spectral line information makes the derived $f_{v,H\alpha}$ a lower limit, however, we demonstrate that the final $f_{v,H\alpha}$ versus redshift relation is close to the true relation (i.e. the true relation here is that of $f_{v,r}$) as there are only a few galaxies in a given redshift slice without this information. Note that as we use all galaxies to derive $f_{v,r}$, the lack of spectral line measurements does not affect $f_{v,r}$ estimates.

whole luminosity function. In appendix A, we describe the formulation of the SWML estimator for bivariate functions. A comparison between the lowest redshift bivariate $L_{H\alpha}$ – M_r function constructed using SWML method and that constructed using the density corrected V_{\max} method is presented in § 3.3.

3.3 A comparison of bivariate LF estimators

The formulation of the classical $1/V_{\max}$ and the density corrected $1/V_{\max}$ is described in § 3.1 and § 3.2. We present the residual

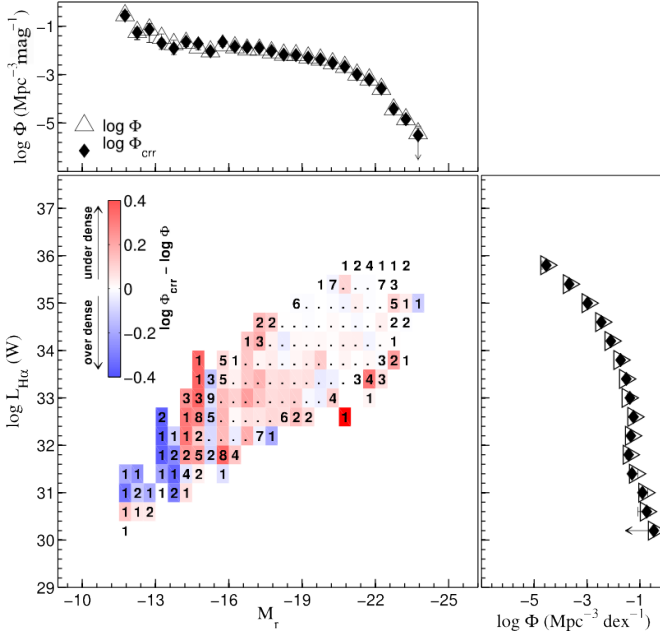


Figure 5. The residual map showing the difference in number densities derived using the classical (Φ) and the density corrected (Φ_{crr}) V_{\max} methods for the $z < 0.1$ bivariate functions. The colours indicate the difference between Φ_{crr} and Φ . A positive difference (redder colours) implies that f_v corrects V_{\max} for an under-density and negative (bluer colours) means the correction is for an over-density. Top and right panels show the M_r and $H\alpha$ univariate LFs.

between the two bivariate functions. We focus here on the $z < 0.1$ slice as this is the one which is most likely to differ between the two methods due to the size of the volume surveyed.

Figure 5 shows the residual maps obtained by subtracting the bivariate $L_{H\alpha}$ – M_r function derived using the density corrected $1/V_{\max}$ method from that derived using the classical $1/V_{\max}$ method.

As mentioned above, the bivariate functions based on the classical $1/V_{\max}$ method, especially the Φ estimates at faint M_r and $L_{H\alpha}$, are affected by large scale structure. The density corrected $1/V_{\max}$ technique is designed to account for radial variations through ρ_{ddp} defined in § 3.2. This is demonstrated in Figure 5, where the redder colour indicates that the applied density weight corrects the classical $1/V_{\max}$ for an under-density and vice versa, with the colour intensity showing the significance of that correction for each $L_{H\alpha}$ – M_r bin. As expected the density correction becomes progressively more important towards fainter M_r and $L_{H\alpha}$ values (i.e. towards increasingly smaller volumes). The top and right panels of Figure 5 show the M_r and $H\alpha$ LFs obtained from integrating the bivariate $L_{H\alpha}$ – M_r functions based on the classical (open symbols) and density corrected (filled symbols) $1/V_{\max}$ methods in $L_{H\alpha}$ and M_r directions, respectively. Clearly, the density corrections to

the bivariate functions have a small effect, and are limited to the faintest end of the bivariate function. This is not surprising given the low volume being probed combined with the small numbers of galaxies contributing to each bin. Higher- z residual maps are devoid of significant differences in this comparison. We note that we see an almost identical result if we use the $f_{V,r}$ rather than $f_{V,H\alpha}$ in making the density-dependent correction to the $V_{\max,H\alpha}$. In summary, we see a marginal difference in the bivariate functions between $1/V_{\max}$ and the density corrected $1/V_{\max}$, while no statistical difference is observed in the univariate LFs.

A similar conclusion is reached from the comparison between SWML with the density corrected V_{\max} (Appendix A).

4 BIVARIATE $L_{H\alpha}$ – M_r FUNCTIONS

The GAMA bivariate $L_{H\alpha}$ – M_r functions in redshift bins constructed using the classical V_{\max} method are shown in Figure 6. Due to the magnitude limited nature of the survey, the range in M_r and $L_{H\alpha}$ probed by the bivariate functions progressively decreases with increasing redshift. Particularly, the lowest- z bivariate LF width with respect to M_r and $L_{H\alpha}$ (i.e. the horizontal and vertical lengths of the bivariate distribution as a function of M_r or $L_{H\alpha}$ respectively) indicates an overall decrease towards fainter $L_{H\alpha}$ – M_r . This is supported by the reduction in number towards fainter M_r and $L_{H\alpha}$. The decrease in horizontal width with increasing $L_{H\alpha}$ is likely to be a result of our sample being biased against red star formers (§ 2). The range in $L_{H\alpha}$ – M_r probed by the three higher- z bivariate functions become more limited and faint $L_{H\alpha}$ – M_r bins become more incomplete with increasing redshift. The $L_{H\alpha}$ – M_r bins with low galaxy number statistics are indicated in Figure 6. Note that the errors in $\log \Phi$ for $L_{H\alpha}$ – M_r bins with small numbers of galaxies (e.g. numbers $\lesssim 3$) is large.

We further assign a photometric blue or red class to each star forming galaxy in our sample. For the analysis presented in this section, the blue–red classification is determined using the $g - r$ and $M_r^{0.1}$ colour–magnitude cut defined in Zehavi et al. (2011) and used by Loveday et al. (2012).

$$(g - r)_{model}^{0.1} = 0.15 - 0.03 M_r^{0.1}. \quad (4)$$

There are a number of blue–red galaxy classification schemes in the literature (e.g. Bell et al. 2003; Baldry et al. 2004; Peng et al. 2010, Taylor et al. submitted). Baldry et al. (2004, 2012), for example, advocate a non-linear cut in $(u - r)$ rest-frame colour and M_r . Since the observed bivariate data distribution in colour–magnitude plane is non-linear (Baldry et al. 2004), a non-linear cut in $(u - r)$ – M_r space would improve our blue–red selection. As our goal in this part of the analysis is to compare the univariate functions computed from the bivariate functions with the univariate functions of Loveday et al. (2012), the same colour–magnitude selection employed by Loveday et al. (2012) is used. A comprehensive discussion of different colour–magnitude classifications and their implications is presented in Taylor et al., (submitted).

The bivariate $L_{H\alpha}$ – M_r functions of blue and red $H\alpha$ SF sub-populations split in four redshift bins are shown in Figure 7. The range in M_r and $L_{H\alpha}$ probed by the blue $L_{H\alpha}$ – M_r functions and their number densities are similar to that of the total $H\alpha$ SF LFs (Figure 6) at all redshifts, implying that the star forming bivariate LF, while drawn from both photometrically classified blue and red sub-populations, consists mostly of blue galaxies.

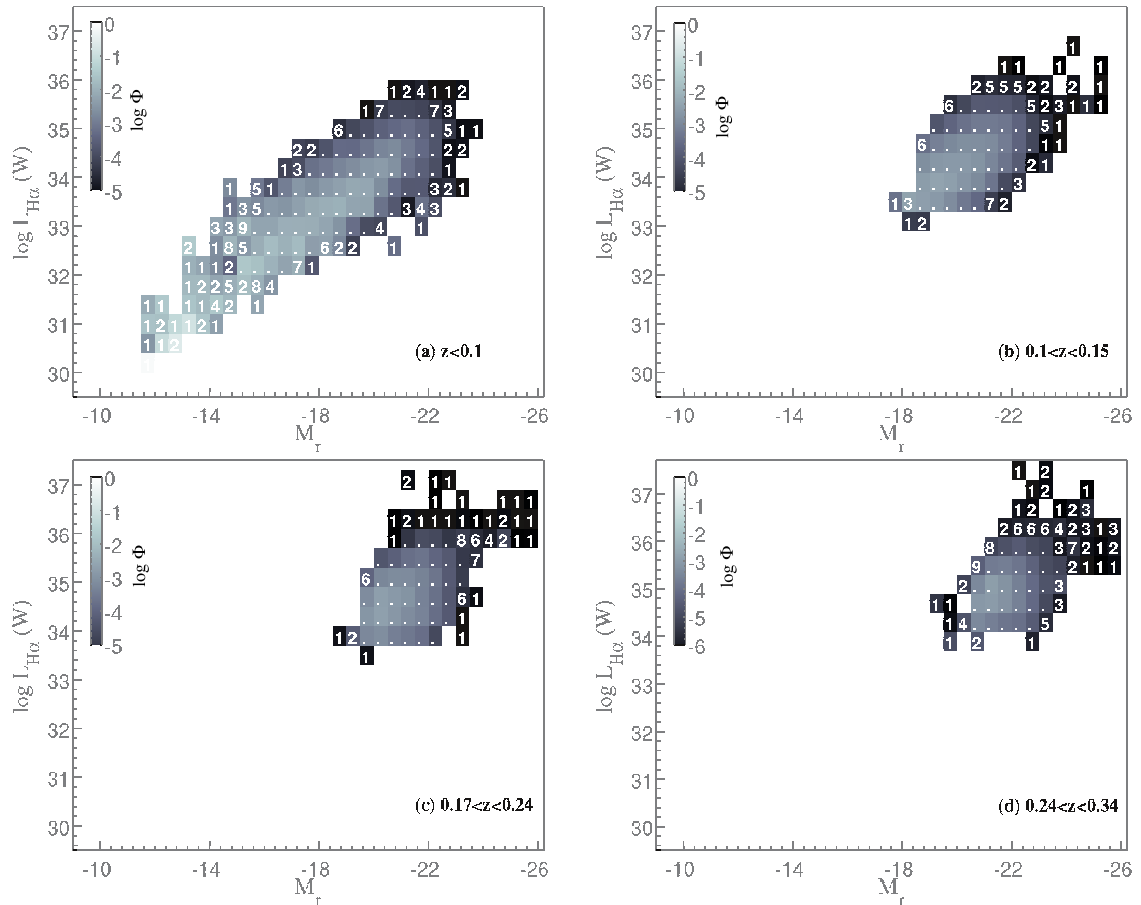


Figure 6. Bivariate $L_{H\alpha}$ – M_r functions of the $H\alpha$ SF galaxies computed using the classical $1/V_{\max}$ method split into four redshift bins with redshift increasing across. The $L_{H\alpha}$ – M_r bins with less than 10 galaxies in them have their number of galaxies quoted, while the dots indicate bins with 10 or more galaxies. The gray scale corresponds to the number density in the unit of $\text{Mpc}^{-3} \text{dex}^{-1} \text{mag}^{-1}$.

4.1 r -band galaxy LFs of $H\alpha$ star formers

By integrating the bivariate $L_{H\alpha}$ – M_r functions over the $L_{H\alpha}$ axis, the M_r LFs of the total $H\alpha$ SF sample and photometrically classified blue and red $H\alpha$ SF sub-samples can be recovered. We explore the evolution of luminosity densities (LDs) in § 5.2.

Figure 8 shows the $z < 0.1$ r -band galaxy LFs of all $H\alpha$ SF (left panel), blue $H\alpha$ SF (right top panel) and red $H\alpha$ SF (right bottom) galaxies computed from integrating the corresponding bivariate functions⁴. These LFs are compared with the GAMA r -band LFs from Loveday et al. (2012) and SDSS r -band LFs from Blanton et al. (2005), which are shown as open squares and diamonds, respectively, in Figure 8.

The $z < 0.1$ $M_r^{0.1}$ LF of all $H\alpha$ SF galaxies (left panel of Figure 8) closely follows the $M_r^{0.1}$ LF of all GAMA galaxies from Loveday et al. (2012). In fact, the bright end of the all $H\alpha$ SF $M_r^{0.1}$ LF very closely tracks the GAMA red $M_r^{0.1}$ LF (Loveday et al. 2012), demonstrating that a significant fraction of the $z < 0.1$ galaxies classified as red have detected $H\alpha$ emission at all $M_r^{0.1}$ values. This is further corroborated by the colour–magnitude distributions (Figure 8 inset) of all $z < 0.1$ galaxies regardless of star formation (filled contours) and $z < 0.1$ $H\alpha$ SF galaxies (purple contours),

where the purple contours extend beyond the blue population. The small discrepancy evident between the blue $M_r^{0.1}$ LF of $H\alpha$ SF galaxies and that from Loveday et al. (2012) is likely a result of the differences in the formulation of the $1/V_{\max}$ technique. By taking into account the minimum of $V_{\max,r}$ and $V_{\max,H\alpha}$ (paper I) rather than only $V_{\max,r}$ as done by Loveday et al. (2012), where we take into account the maximum volume corrections based on both r -band magnitude and $H\alpha$ flux limits.

Furthermore, we use the Kauffmann et al. (2003) criteria on discriminating pure-SF and SF-AGN composites to further remove likely SF-AGN composites from the red SF sample. The resultant $M_r^{0.1}$ LF is shown as open red diamonds in the right bottom panel of Figure 8. The open symbols are slightly displaced from the $M_r^{0.1}$ LF constructed using SF galaxies selected based on the Kewley & Dopita (2002) criteria (filled symbols) at the bright end of the LF, and overlap with the filled symbols at the faint-end, implying that only a small fraction of red galaxies are removed from the original red SF sample by the Kauffmann et al. (2003) cut. We have also used the method of Cid Fernandes et al. (2011) for differentiating SF galaxies from AGNs. Even with the AGN/SF cuts recommended by Cid Fernandes et al. (2011), more than 50% of the red galaxy sample at a given redshift retain their star forming status. So that a fraction of galaxies classified as red at all $M_r^{0.1}$ values are in fact currently forming stars. Moreover, the analysis of Lara-López et al. (2013) exploring the properties of SDSS and GAMA galaxies find that a large fraction of galaxies detected in GAMA are SF in comparison to SDSS as GAMA is deeper than

⁴ The bivariate LFs shown in Figures 6 and 7 use r -band absolute magnitude k -corrected to $z = 0$. To obtain the $M_r^{0.1}$ LFs shown in Figure 8, we calculate the bivariate LFs using r -band absolute magnitude k -corrected to $z = 0.1$ to compare our results directly with Loveday et al. (2012).

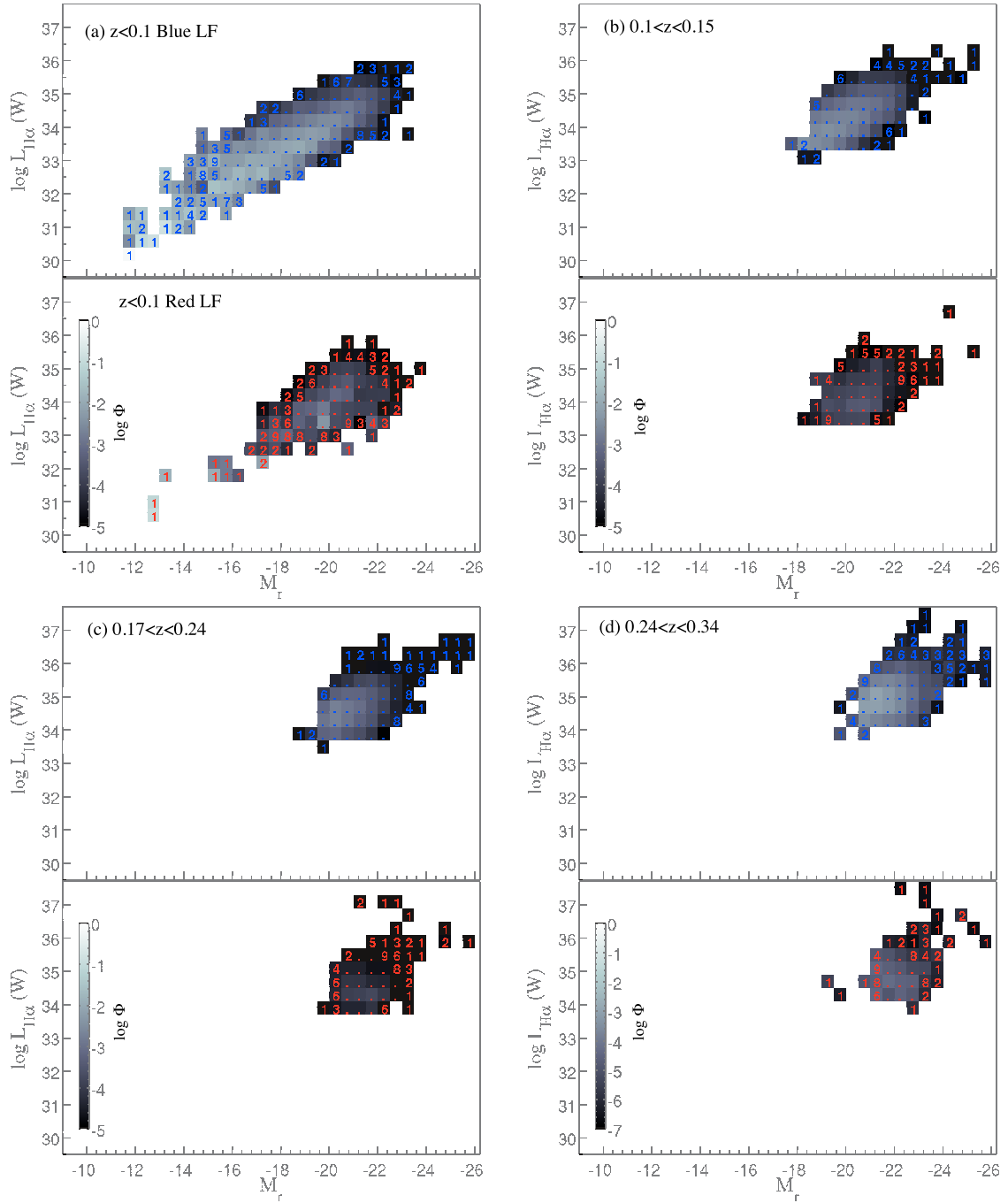


Figure 7. Bivariate $L_{H\alpha}$ – M_r functions of blue and red $H\alpha$ SF sub-populations constructed using the classical V_{\max} method split in four redshift bins. The gray scale shown along side each red bivariate LF indicates the number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-1} \text{mag}^{-1}$ for both blue and red functions of the respective redshift range. Like in Figure 6, the $L_{H\alpha}$ – M_r bins with less than 10 galaxies in have their number of galaxies indicated.

SDSS, and therefore more sensitive to low mass galaxies at low redshift, which are mostly dominated by star formation.

The higher- z (z up to 0.34) $M_r^{0.1}$ LF of all $H\alpha$ SF galaxies (Figure B2) overlap with the GAMA blue LFs of Loveday et al. (2012) over similar redshift ranges, while the fractional contribution from red $H\alpha$ SF galaxies to the GAMA red $M_r^{0.1}$ LFs (Loveday et al. 2012) progressively drops with increasing redshift. The contours in the insets of Figure 9 depicting the distribution of all galaxies (filled contours) and SF galaxies (magenta contours) at different redshifts show that the contours of SF galaxies become more restricted to the blue sub-population with increasing redshift, corroborating the drop in red SF number densities seen in Figure B2. This is most likely due to the flux limit biasing our sample against

these systems (Figure 1), and possibly also due to the difficulty in measuring spectral lines in more obscured, weak emission line systems at higher redshifts.

5 MODELLING OF BIVARIATE $L_{H\alpha}$ DEPENDENT FUNCTIONS AND IMPLICATIONS FOR LUMINOSITY AND SFR DENSITIES

LFs, emission-line and photometric alike, are traditionally parameterised with a Schechter (1976) function, which is then integrated to obtain a luminosity or SFR density. The linear form of the

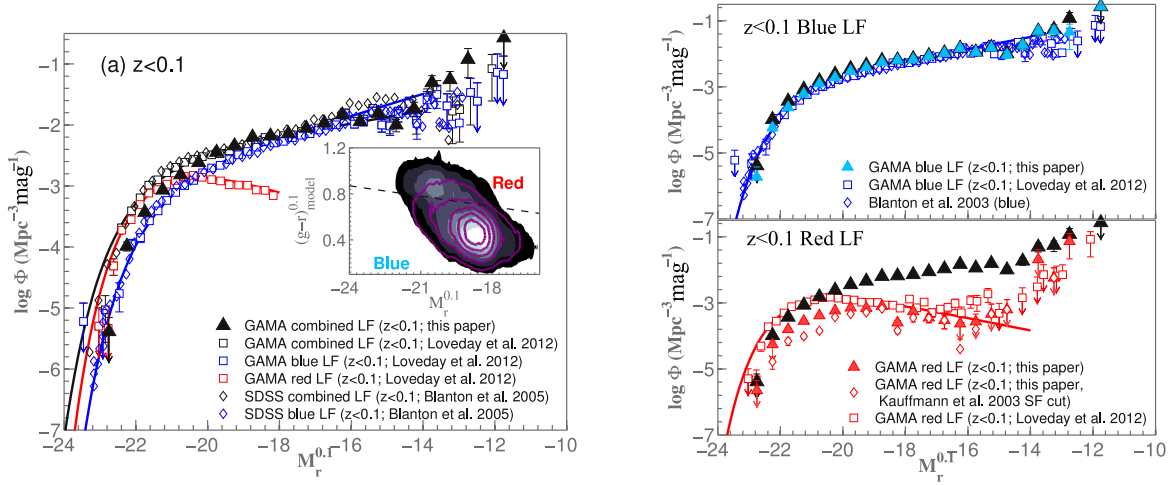


Figure 8. GAMA M_r total (left panel), blue (right top) and red (right bottom) LFs of star forming galaxies derived from integrating the $z < 0.1$ bivariate $L_{H\alpha}$ – M_r functions. These LFs are compared against the M_r LFs presented in Loveday et al. (2012) for all GAMA galaxies (open squares) and Blanton et al. (2005) for all SDSS galaxies (open diamonds). The solid lines show the best fitting Schechter functions to $z < 0.1$ LFs from Loveday et al. (2012). The left panel inset shows the data density distributions in linear space of all $z < 0.1$ galaxies regardless star formation (filled contours) and star forming galaxies with $F_{H\alpha} > 1 \times 10^{-18} \text{ W/m}^2$ (solid contours) in $(g-r)^{0.1}_{\text{model}}$ and $M_r^{0.1}$ plane. The colour–magnitude cut (Eq. 4) used to split blue and red galaxies is shown as a dashed line. The open red diamonds in right bottom panel indicate the red SF $M_r^{0.1}$ LF constructed using the (pure) SF galaxies selected according to Kauffmann et al. (2003) SF/AGN prescription instead of that of Kewley & Dopita (2002).

Schechter function⁵,

$$\Phi(L)dL = \Phi^* \left(\frac{L}{L^*} \right)^\alpha \exp \left(-\frac{L}{L^*} \right) d \left(\frac{L}{L^*} \right), \quad (5)$$

behaves as a power–law with a slope α for luminosities (L) less than the characteristic luminosity (L^*) and as an exponential for $L > L^*$, with the normalisation given by Φ^* . From this, the predicted luminosity density is given by,

$$\rho_{L,fit} = \int_0^\infty L \Phi(L) dL = \Phi^* L^* \Gamma(\alpha + 2), \quad (6)$$

where Γ is the Gamma function, and a conservative limit is given by,

$$\rho_{L,fit,lim} = \int_0^{L_{min}} L \Phi(L) dL = \Phi^* L^* \Gamma(\alpha + 2, L_{min}/L^*). \quad (7)$$

where Γ is now the incomplete Gamma function.

Galaxy broadband LFs are, generally, well described by a Schechter function (e.g. Hill et al. 2010; Loveday et al. 2012). Several studies (e.g. Blanton et al. 2005), however, find that a double Schechter function is best suited at capturing the whole shape of the galaxy LF, especially if the range in magnitude probed is relatively large.

While the galaxy broadband LFs show an exponential–like fall in Φ with increasing brightness, the star–forming LFs (e.g. $H\alpha$, far infrared, radio) show a less steep fall in Φ (Saunders et al. 1990; Salim & Lee 2012, paper I). For this reason, to characterise the star forming LFs we adopt a Saunders et al. (1990) function. The

linear form of the Saunders function,

$$\Phi(L)dL = \Phi^* \left(\frac{L}{L^*} \right)^\alpha \exp \left[-\frac{1}{2\sigma^2} \log^2 \left(1 + \frac{L}{L^*} \right) \right] d \left(\frac{L}{L^*} \right), \quad (8)$$

behaves in a similar fashion to a Schechter function for $L < L^*$ and as a Gaussian in log luminosity with a width (σ) for $L > L^*$.

We fit Schechter functions to the $M_r^{0.1}$ LFs of SF galaxies shown in Figures 8 and B2 using a Levenberg–Marquardt routine to find the minimum χ^2 to the binned LF data points. The resultant Schechter functional fits and their best fitting parameters are presented in Figure 10 and Table 1. Due to the lack of faint SF galaxies at higher– z , a consequence of the survey magnitude selection, the best fitting slopes for the $0.1 < z < 0.15$, $0.17 < z < 0.24$ and $0.24 < z < 0.34$ LFs cannot be measured reliably from the observed LF. Instead, we constrain the faint–end slopes of higher– z LFs to be equal to the best fitting slopes of their respective $z < 0.1$ LFs. The low– z red SF galaxy LF has a poorly constrained α due to a lack of bright galaxies necessary to disentangle the degeneracy between α and M^* , therefore we assume $\alpha = -1.29$ (i.e. α estimated from the $z < 0.1$ $M_r^{0.1}$ LF all $H\alpha$ SF galaxies) for the higher– z red LFs.

5.1 Functional fits to bivariate functions

In this section, we explore a simple analytic approach to modelling the bivariate $L_{H\alpha}$ – M_r function. Such a parameterisation of bivariate functions is useful in comparing the distributions drawn from differently selected samples and for studying the redshift evolution inclusive of selection biases.

Choloniewski (1985) and de Jong & Lacey (2000) developed a formalism to link the distribution of galaxy scale sizes, assumed to be Gaussian, to the luminosity parameterised by a Schechter function. Their analytic expression as well as other related functional forms have widely been used to model bivariate brightness profiles (e.g. de Jong & Lacey 2000; Cross et al. 2001; Driver et al. 2005; Ball et al. 2006a), size–luminosity (e.g. de Jong & Lacey 2000; Huang

⁵ The logarithmic form of the Schechter function is expressed as,

$$\Phi(\log L)d \log L = \ln(10)\Phi^* 10^{(\log L - \log L^*)(\alpha+1)} \exp[-10^{(\log L - \log L^*)}] d \log L.$$

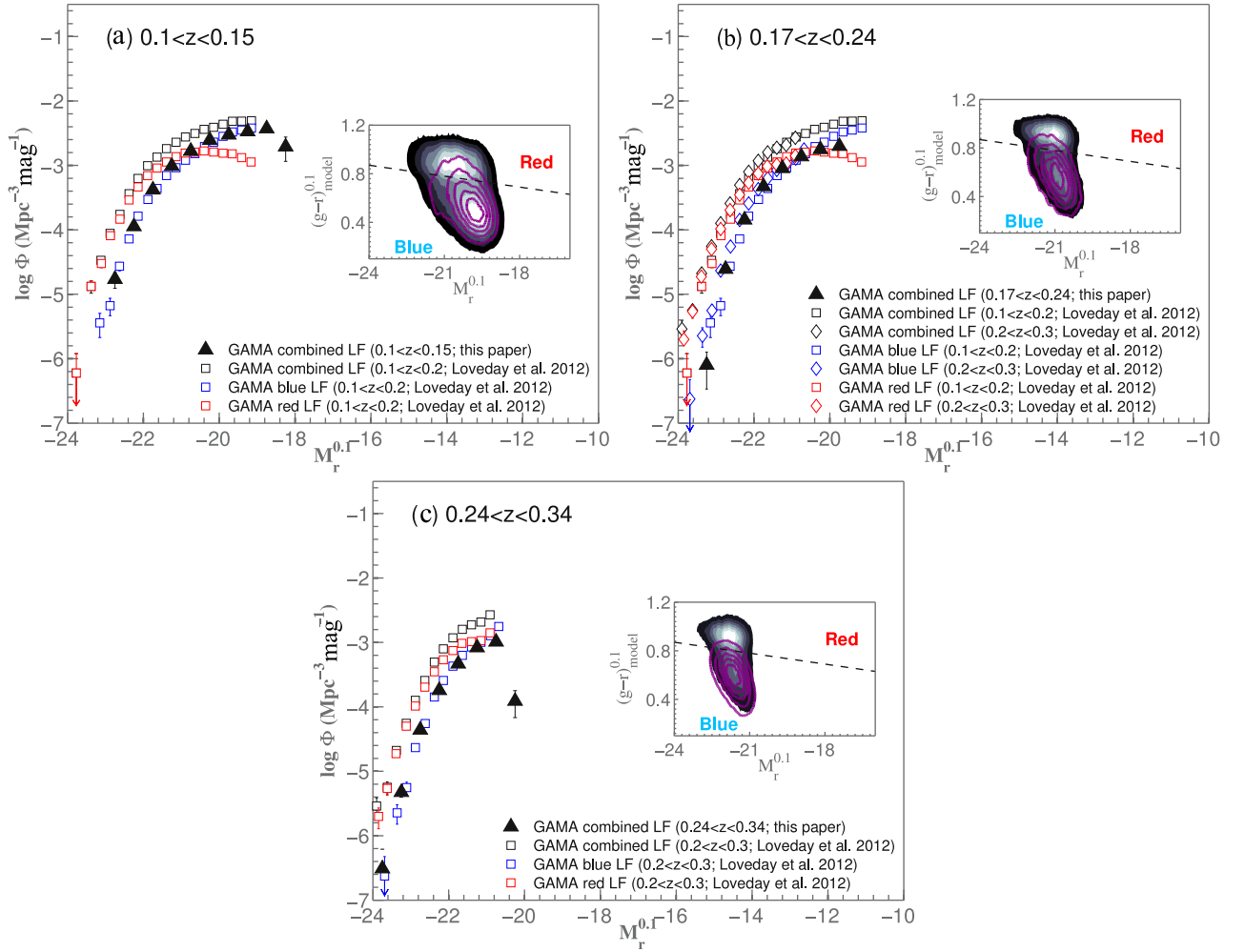


Figure 9. GAMA higher- z $M_r^{0.1}$ LFs of all SF galaxies computed by integrating the bivariate $L_{H\alpha}$ - M_r functions. The open symbols in each panel show the GAMA all, blue and red $M_r^{0.1}$ LFs of Loveday et al. (2012) over similar redshift ranges. The insets show the data density distributions in $(g-r)_{model}^{0.1}$ and $M_r^{0.1}$ of all galaxies regardless of star formation (grey shading) and SF galaxies with $F_{H\alpha} > 1 \times 10^{-18}$ W/m² (magenta contours). These distributions correspond to the redshift ranges given in the plot key. The colour-magnitude cut (Eq. 4) is shown as a dashed line.

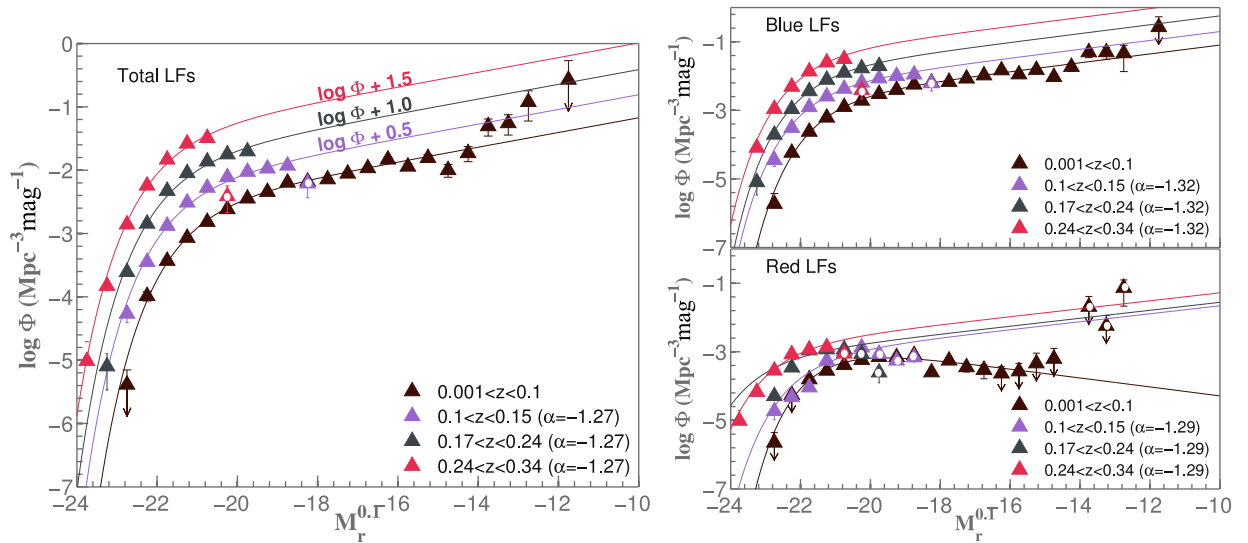


Figure 10. The best fitting Schechter functions to the $M_r^{0.1}$ LFs shown in Figures 8, 9 and B2. The faint-end slope, α , is fixed at -1.29 for the higher- z combined (left panel) and red SF (right bottom) LFs. It is fixed at -1.32 for the higher- z blue LFs (Right top). Each higher- z LF is scaled up by the factors stated in the left panel to aid legibility.

Table 1. The best fitting Schechter function parameters corresponding to the functional fits shown in Figure 10, and the luminosity densities computed from integrating the fits.

z	M_r^*	$\log \Phi^*$ (Mpc^{-3})	α	$\log \rho_{Lfit}$ ($L_\odot \text{ Mpc}^{-3}$)
Combined				
$z < 0.10$	-20.93 ± 0.15	-2.41 ± 0.14	-1.29 ± 0.06	7.97 ± 0.05
$0.10 < z < 0.15$	-21.21 ± 0.06	-2.54 ± 0.05	-1.29^a	7.96 ± 0.03
$0.17 < z < 0.24$	-21.34 ± 0.13	-2.64 ± 0.15	-1.29^a	7.91 ± 0.09
$0.24 < z < 0.34$	-21.52 ± 0.02	-2.72 ± 0.04	-1.29^a	7.90 ± 0.03
Blue				
$z < 0.10$	-20.80 ± 0.07	-2.46 ± 0.07	-1.32 ± 0.05	7.89 ± 0.01
$0.10 < z < 0.15$	-21.21 ± 0.08	-2.62 ± 0.06	-1.32^b	7.89 ± 0.06
$0.17 < z < 0.24$	-21.25 ± 0.07	-2.66 ± 0.11	-1.32^b	7.86 ± 0.08
$0.24 < z < 0.34$	-21.40 ± 0.06	-2.71 ± 0.08	-1.32^b	7.89 ± 0.05
Red				
$z < 0.10$	-20.76 ± 0.09	-2.85 ± 0.08	-0.67 ± 0.07	7.31 ± 0.04
$0.10 < z < 0.15$	-21.44 ± 0.25	-3.45 ± 0.17	-1.29^a	7.14 ± 0.06
$0.17 < z < 0.24$	-22.74 ± 1.07	-4.00 ± 0.27	-1.29^a	7.12 ± 0.16
$0.24 < z < 0.34$	-22.16 ± 0.17	-4.16 ± 0.22	-1.29^a	6.82 ± 0.14

^a α is fixed at -1.29 , the best fitting slope for the $z < 0.1$ blue and red combined $M_r^{0.1}$ LF of H α SF galaxies.

^b α is fixed at -1.32 , the best fitting slope for the $z < 0.1$ blue $M_r^{0.1}$ LF of H α SF galaxies.

et al. 2013) and colour–luminosity relationships (e.g. Chapman et al. 2003; Chapin et al. 2009). Similar functional forms have also been formulated by Yang et al. (2005) and Cooray & Milosavljević (2005) in the context of conditional LFs that specify the average number of galaxies with luminosities in the range $L \pm \Delta L/2$ that reside in a given halo mass.

Our motivation for modelling the GAMA bivariate functions is to correct for the apparent lack of evolution in H α SFR densities with redshift reported in paper I. The low redshift bivariate model can be used as reference to gain an understanding of the extent to which the higher- z ($0.1 \lesssim z \lesssim 0.34$) samples are affected by incompleteness.

Below we describe the construction of a simple analytic model, denoted $\Psi(M, L_{H\alpha})$, to describe the bivariate functions presented in this paper, assuming that the bivariate functions can be written as a product of two functions (Choloniewski 1985; Corbelli et al. 1991; Hopkins 1998). Naturally, as a Schechter function best describes the M_r LFs presented in §4.1 and a Saunders function best describes the GAMA H α LFs (paper I), these functions are adopted to represent the bivariate distributions.

To link these functional forms in a bivariate analytic relation, we begin by establishing how $L_{H\alpha}^*$ in the H α LF varies as a function of M_r . The relationship between $L_{H\alpha}^*$ and M_r is clearly evident in Figure 11, where we divide the lowest- z $L_{H\alpha}$ – M_r function by $L_{H\alpha}$ LF at a given M_r . The symbols indicate the best fitting $L_{H\alpha}^*$ in different magnitude ranges, estimated by fitting a Saunders function (Eq. 8) with a non-varying faint-end slope (blue) to the data and fitting a Gaussian (green),

$$\Phi(L)dL = \frac{\Phi^*}{\sigma\sqrt{2\pi}} \exp\left[-\frac{1}{2\sigma^2} \log^2\left(\frac{L}{L^*}\right)\right] d\left(\frac{L}{L^*}\right), \quad (9)$$

with a Gaussian width, σ , that is allowed to vary to the data.

The relationship between $L_{H\alpha}^*$ and M_r seen in Figure 11 can be approximated as

$$L_{H\alpha}^* = L_{H\alpha}^*(M_r^0) 10^{-0.4\beta(M_r - M_r^0)}. \quad (10)$$

Using Eq. 10 to connect Eq. 5 and 8, and expressing the distribu-

tion in terms of M_r , we arrive at the full bivariate expression:

$$\begin{aligned} \Psi(M_r, L_{H\alpha}) &= \Phi(M_r) \times \Phi(M_r, L_{H\alpha}) = 0.4 \ln(10) \\ &\Psi^* 10^{-0.4(\alpha_M + 1)(M_r - M_r^*)} \\ &\times \exp[-10^{-0.4(M_r - M_r^*)}] \\ &\times \left\{ \left[\frac{L_{H\alpha}}{L_{H\alpha}^*(M_r^0)} \right] \left[10^{0.4\beta(M_r - M_r^0)} \right] \right\}^{\alpha_L} \\ &\times \exp\left\{ -\frac{1}{2\sigma^2} \log^2 \left[1 + \left(\frac{L_{H\alpha}}{L_{H\alpha}^*(M_r^0)} \right) \left(10^{0.4\beta(M_r - M_r^0)} \right) \right] \right\}. \end{aligned} \quad (11)$$

We assume $M_r^0 = -19.0$ and fit the, now, seven free parameter model to the $z < 0.1$ $L_{H\alpha}$ – M_r function through a nonlinear χ^2 minimisation routine based on Levenberg–Marquardt method using the Poisson errors of the bivariate function measurements. The resultant best fitting model is shown in Figure 12 and the best fitting model parameters are given in Table 2.

To determine the best fitting parameters for the higher- z ($0.1 \lesssim z \lesssim 0.34$) bivariate functions, we fix α_M and α_L to be equal to the best-fitting $z < 0.1$ model values. The best fitting values determined in this manner are given in Table 2. Given the log-normal like shape of the $z < 0.1$ normalised bivariate function at brighter magnitudes (Figure 11), which is similar to that exhibited by size-magnitude or colour-magnitude distributions (Choloniewski 1985; Chapman et al. 2003; Huang et al. 2013), a Schechter–Gaussian model can also be fitted.

$$\begin{aligned} \Psi(M_r, L_{H\alpha}) &= 0.4 \ln(10) \\ &\Psi^* 10^{-0.4(\alpha_M + 1)(M_r - M_r^*)} \\ &\times \exp[-10^{-0.4(M_r - M_r^*)}] \\ &\times \exp\left\{ -\frac{1}{2\sigma^2} \log^2 \left[1 + \left(\frac{L_{H\alpha}}{L_{H\alpha}^*(M_r^0)} \right) \left(10^{0.4\beta(M_r - M_r^0)} \right) \right] \right\}. \end{aligned} \quad (12)$$

The number density contours produced by a Schechter–Gaussian model is identical to the Φ contours shown in Figure 12. Therefore in Table 2 we also provide the best-fitting Schechter–Gaussian model parameters for the bivariate functions shown in Figure 6. We note that in the two highest redshift bins, we have a limited sampling of galaxies fainter than broadband or emission line characteristic luminosity (Figure 6).

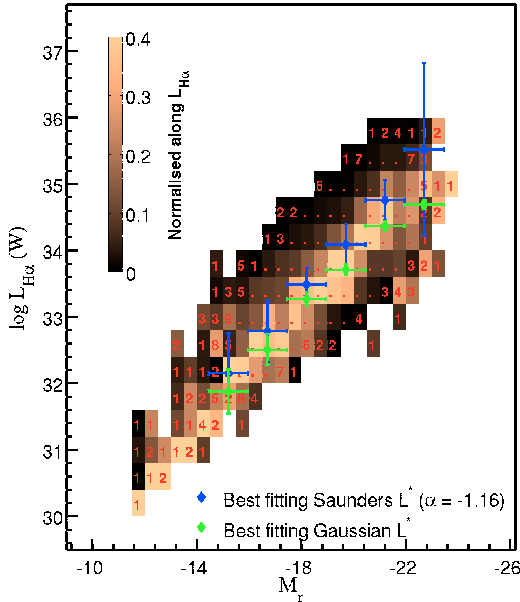
Table 2. The best fitting parameters of the fitted models (Eq. 11) to the four bivariate $L_{H\alpha}$ – M_r functions. The faint-end slopes, α_M and α_L , of the higher- z bivariate models are fixed to be equal to the $z < 0.1$ model values.

Parameter	Schechter–Saunders function			
	$0.001 < z < 0.1$	$0.1 < z < 0.15$	$0.17 < z < 0.24$	$0.24 < z < 0.34$
M_r^*	-21.12 ± 0.10	-21.57 ± 0.09	-21.77 ± 0.26	-22.62 ± 0.14
α_M	-1.29 ± 0.06	-	-	-
$\log \Psi^* (\text{Mpc}^{-3})$	-4.17 ± 1.54	-4.49 ± 0.45	-4.79 ± 0.49	-4.40 ± 0.93
$\log L_{H\alpha}^* (W)$	32.30 ± 0.65	32.52 ± 0.33	32.75 ± 0.59	33.55 ± 0.64
β	0.92 ± 0.03	0.85 ± 0.12	0.59 ± 0.30	0.46 ± 0.25
σ	0.46 ± 0.03	0.46 ± 0.05	0.48 ± 0.04	0.38 ± 0.10
α_L	1.53 ± 1.15	-	-	-

Parameter	Schechter–Gaussian function			
	$0.001 < z < 0.1$	$0.1 < z < 0.15$	$0.17 < z < 0.24$	$0.24 < z < 0.34$
M_r^*	-21.12 ± 0.11	-21.57 ± 0.10	-21.77 ± 0.27	-22.63 ± 0.14
α_M	-1.29 ± 0.06	-	-	-
$\log \Psi^* (\text{Mpc}^{-3})$	-2.67 ± 0.14	-2.94 ± 0.14	-3.08 ± 0.33	-3.53 ± 0.31
$\log L_{H\alpha}^* (W)$	33.55 ± 0.04	33.79 ± 0.14	34.12 ± 0.42	34.45 ± 0.36
β	0.92 ± 0.03	0.85 ± 0.12	0.59 ± 0.30	0.44 ± 0.24
σ	0.45 ± 0.02	0.45 ± 0.04	0.47 ± 0.04	0.38 ± 0.07

Table 3. The SFRDs estimated from integrating the best-fitting Schechter–Saunders models over three different $L_{H\alpha}$ and M_r ranges. The limiting $\log L_{H\alpha} = 37$ (denoted L_b in the table) and $M_r = -26$ (denoted M_b in the table). The second table column ($\log \dot{\rho}_*(0,0)-(L_b, M_b)$) reports the SFRDs computed from integrating the best-fitting higher- z bivariate models from $[L_{H\alpha}, M_r] = [0, 0]$ to $[L_b, M_b]$, the third provides those computed from integrating the bivariate models between the observed lowest $z < 0.1$ $\log L_{H\alpha}$ and M_r values and $[L_b, M_b]$, and the final column provides the SFRDs obtained from integrating each model between the lowest $\log L_{H\alpha}$ and M_r values at their respective redshift range and $[L_b, M_b]$.

Redshift range	$\log \dot{\rho}_*(0,0)-(L_b, M_b)$ ($\text{M}\odot\text{yr}^{-1}\text{Mpc}^{-3}$)	$\log \dot{\rho}_*(L_{f,z1}, M_{f,z1})-(L_b, M_b)$ ($\text{M}\odot\text{yr}^{-1}\text{Mpc}^{-3}$)	$\log \dot{\rho}_*(L_{f,zx}, M_{f,zx})-(L_b, M_b)$ ($\text{M}\odot\text{yr}^{-1}\text{Mpc}^{-3}$)
$0.1 < z < 0.15$	-2.00	-2.00	-2.03 (-14.25, 33.1)
$0.17 < z < 0.24$	-1.75	-1.77	-2.03 (-18.75, 33.5)
$0.24 < z < 0.34$	-1.72	-1.77	-2.14 (-19.25, 33.9)

**Figure 11.** The low- z bivariate function (Figure 6a) normalised along the $L_{H\alpha}$ direction. The symbols indicate the variation in characteristic luminosity ($L_{H\alpha}^*$) with respect to M_r for Saunders functional fits (blue) and Gaussian fits (green) to the $H\alpha$ LFs in different M_r ranges. For the Saunders fits, the faint-end slope (α) is fixed at -1.16 , the best fitting α of the $z < 0.1$ GAMA $H\alpha$ LF (paper I). The horizontal errors indicate the M_r range probed and the vertical errors indicate the errors associated with best fitting $L_{H\alpha}^*$ values. In both cases, the variation in $L_{H\alpha}^*$ can be approximated through a power-law (Eq. 10).

More complex fitting methods like that of de Jong & Lacey (2000) could be considered, but for the purposes of this analysis, however, we find the best-fit model shown in Figure 12 to be sufficient as it provides a good qualitative and a quantitative description of the low- z bivariate $L_{H\alpha}$ – M_r function. Furthermore, the approximate Schechter and Saunders functional fit forms of GAMA M_r and $H\alpha$ LFs can be recovered from numerically integrating $\Psi(M_r, L_{H\alpha})$ along $L_{H\alpha}$ and M_r axes, respectively. Moreover, by integrating $\Psi(M_r, L_{H\alpha})$ with respect to both $L_{H\alpha}$ and M_r , we recover the $z < 0.1$ $H\alpha$ SFR density reported in paper I.

The best fitting $z < 0.1$ M_r^* and α_M bivariate model parameters given in Table 2 agree within uncertainty with the best fitting parameters determined for the $z < 0.1$ univariate M_r LF (Table 1). The relationship between $\log L_{H\alpha}^*$ and M_r^* is emphasised in Figure 11. Note that α_L represents the faint-end slope, and the positive slope implies a decrease in number density. This is not unexpected as the distribution of the normalised number densities with respect to $\log L_{H\alpha}$ within a given magnitude bin has a log-normal shape than a power-law shape (Figure 11).

In summary, both Schechter–Saunders and Schechter–gaussian models are able to provide a good representation of the lowest- z bivariate $L_{H\alpha}$ – M_r function. As for the higher- z ($0.1 \lesssim z \lesssim 0.34$) bivariate $L_{H\alpha}$ – M_r functions, the models provide a good description of the bright end of the bivariate functions, where the data exists. It becomes increasingly difficult to constrain the models to obtain a good description of the faint-end of the bivariate functions with increasing redshift as the range in $L_{H\alpha}$ and M_r probed decreases.

Finally, an alternative method of modelling bivariate functions is

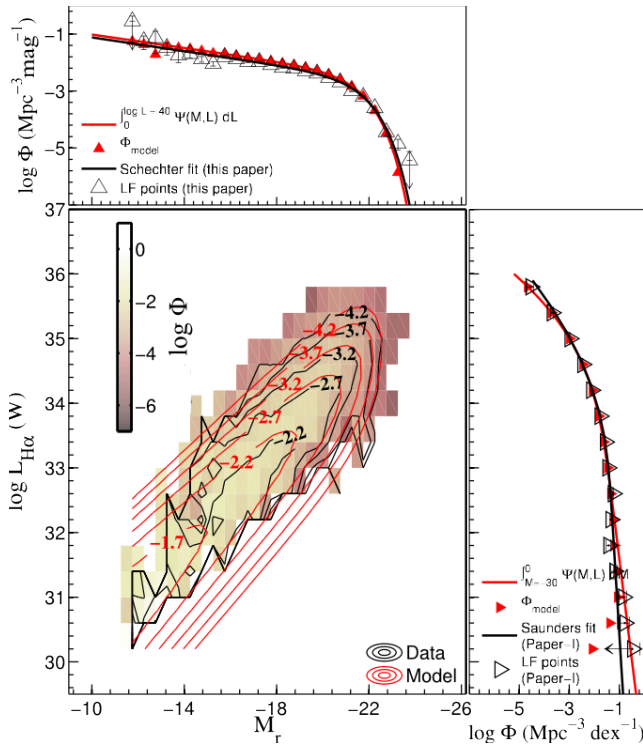


Figure 12. The $z < 0.1$ bivariate function and the fitted model (Eq. 11). The model contours provide a reasonable description of the data (black lines). The top and right panels compare the univariate M_r (Figure 8) and $H\alpha$ (Figure 7 of paper I) LFs (symbols) computed from integrating the $z < 0.1$ bivariate M_r – $H\alpha$ function (Figure 6a) with that predicted by the model. Also shown are the Schechter and Saunders functional fits to the univariate M_r and $H\alpha$ LFs (Figure 11 of paper I). The colour scale indicates log number densities in the unit of $\text{Mpc}^{-3} \text{dex}^{-1} \text{mag}^{-1}$.

proposed by Takeuchi (2010). This approach relies on using a copula to connect two marginal distributions (e.g. $H\alpha$ and M_r LFs) thereby constructing their bivariate distribution. Takeuchi (2010) and Takeuchi et al. (2013) advocate the use of a Farlie–Gumbel–Morgenstern or a Gaussian copula, both of which are explicitly related to the linear correlation coefficient, to connect two given marginals. See Takeuchi (2010) and Sato et al. (2011) for a rigorous mathematical definition of copula theory, dependence measures between two variables and how to estimate the bivariate distribution given two or more marginals.

5.2 Luminosity and density of SF galaxies

In this section we present the r -band luminosity and $H\alpha$ SFR density of SF galaxies computed from integrating the bivariate $L_{H\alpha}$ – M_r functions. The luminosity density (ρ_L) evolution observed in the GAMA $H\alpha$ SF population is shown in Figure 13. The filled diamonds indicate the GAMA ρ_L derived from integrating the four Schechter functions shown in Figure 10. The filled stars indicate the density estimated from integrating the best-fitting Schechter–Saunders model to the $0.1 < z < 0.15$ bivariate $L_{H\alpha}$ – M_r function. The $z < 0.1$ estimate is not shown here as it is similar to that obtained from integrating the Schechter functional fit to the univariate $z < 0.1$ M_r LF. We also show the GAMA ρ_L confidence limits from Loveday et al. (2012), and low redshift density estimates from Blanton et al. (2003b) and Montero-Dorta & Prada (2009). We see a result here that is consistent with the SF populations of the broadband blue and red LFs shown in Figure B2. That is the ρ_L estimated from the best-fitting Schechter functions do not show a evolution in ρ_L with redshift, in contrast to Loveday et al. (2012). This is a natural consequence of the SF populations

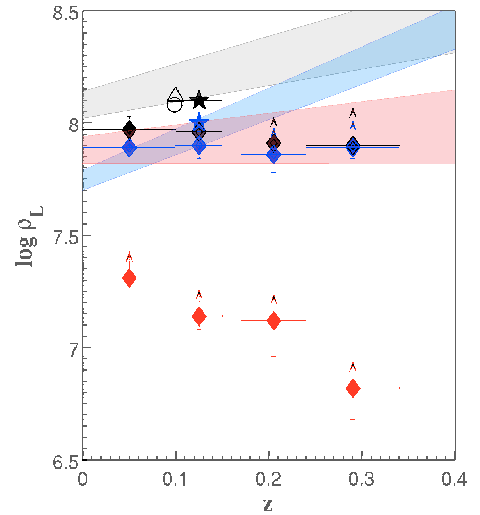


Figure 13. The evolution of r -band luminosity density (ρ_L) of all, and blue and red $H\alpha$ SF galaxies as a function of redshift (filled black, blue and red symbols respectively), compared to the evolution of ρ_L of all, and blue and red GAMA galaxies as a function of redshift (shaded regions; Loveday et al. 2012). Also shown are the ρ_L measurements at low- z from Blanton et al. (2003b) (open circle) and Montero-Dorta & Prada (2009) (open triangle). The filled stars at $z \sim 0.12$ indicate the value obtained from integrating the best-fitting bivariate analytical form for the $0.1 < z < 0.15$ bivariate $L_{H\alpha}$ – M_r function.

comprising only a small fraction (10–20%) of the total red galaxy population. The decrease in ρ_L of blue SF population at higher- z ($z > 0.17$) is likely due to the difficulty in measuring $H\alpha$ in higher- z galaxy spectra as higher- z galaxies likely to have fainter optical magnitudes and lower spectral signal-to-noise than their low- z counterparts. Given our sample is already biased against red SF galaxies, mainly as a result of the $H\alpha$ flux limit, the drop in ρ_L corresponding those galaxies is not unexpected.

5.3 Star Formation Rate Density

As mentioned previously, our primary motivation behind modelling the bivariate $L_{H\alpha}$ – M_r functions is to overcome the bivariate sample effects introduced by the dual $H\alpha$ flux and magnitude selection imposed on our sample. As a result of this effect, the higher- z ($z > 0.1$) SFRDs presented in paper I are underestimates. In Figure 14 we show the SFRDs derived by integrating the bivariate analytic fit to $L_{H\alpha}$ – M_r function. By modelling the low-redshift $L_{H\alpha}$ – M_r distribution over $-26 < M_r < -10$ and $30 < \log L_{H\alpha} < 37$ (Figure 12), and assuming the faint-end bivariate distribution is similar for the higher- z ($0.1 \lesssim z \lesssim 0.34$) samples, we can estimate a correction for the missing optically faint star forming galaxies at those redshifts. In Figure 14 it can be seen that the resulting SFRDs are much more consistent with that from other published measurements (e.g. Hopkins & Beacom 2006), than the direct estimates from paper I. Note that the two highest redshift ranges probed likely overestimated due to poorly constrained bivariate functions.

6 IMPORTANCE OF BIVARIATE $L_{H\alpha}$ – M FUNCTIONS

In this section, we investigate the evolution of the bivariate $L_{H\alpha}$ – M functions, and the SMFs that result from integrating the bivariate functions for the full sample and those split by colour. The

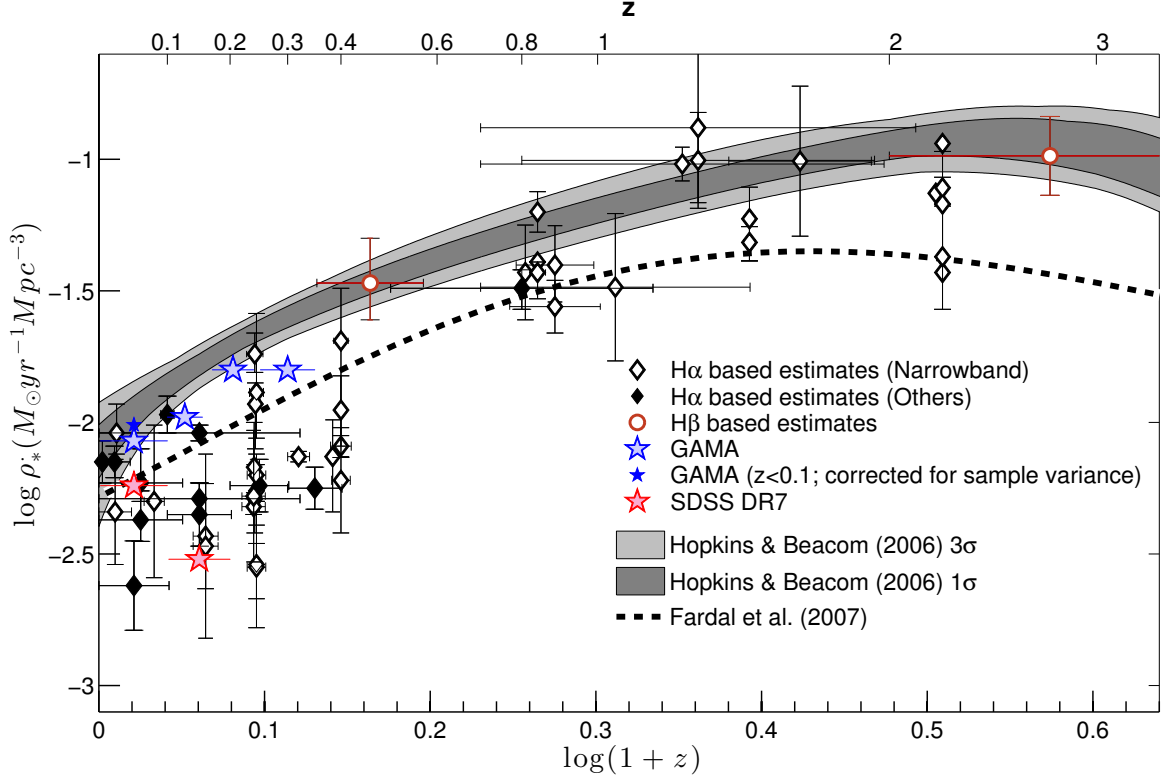


Figure 14. The cosmic SFR history, taken from paper I, with our new measurements shown as blue stars. The GAMA SFRDs are based on integrating the analytic fits to the bivariate $L_{H\alpha}$ – M_r LFs (Figures 12). Published estimates based on narrowband surveys and slitless spectroscopy data are shown as open symbols, and those based on broadband surveys as filled symbols. The grey bands and the dashed line correspond to the best-fitting star formation histories of Hopkins & Beacom (2006) and Fardal et al. (2007), respectively.

bivariate $L_{H\alpha}$ – \mathcal{M} functions of all SF galaxies in redshift bins are shown Figure 16. Again the lowest redshift function probes the largest range in both $L_{H\alpha}$ and \mathcal{M} , while the higher- z ($0.1 < z < 0.34$) functions indicate a decrease in range with increasing redshift like seen in Figure 6 for bivariate $L_{H\alpha}$ – M_r functions..

The colour–magnitude relation used to obtain a photometric blue and red classification in this part of the analysis is different to that discussed in § 4. To reiterate, our objective in constructing bivariate and univariate SMFs of star forming galaxies with blue and red photometric class is to compare our results with the existing GAMA results in a consistent manner (e.g. Baldry et al. 2012). For the stellar mass based analyses presented in the subsequent sections, as they are focussed on SMFs, we use the colour–magnitude relation introduced by Baldry et al. (2004) (Eq. 11 of their paper) and used by Baldry et al. (2012) to construct the $z < 0.06$ GAMA SMFs split by colour.

6.1 The Stellar Mass Functions of SF galaxies

Baldry et al. (2012) present the $z < 0.06$ GAMA galaxy SMFs determined using the density corrected $1/V_{\max}$ method. They further show the GAMA SMFs of photometrically classified blue and red galaxies based on the colour ($u-r$)–magnitude (M_r) relationship given in Baldry et al. (2004). Using the density corrected $1/V_{\max}$ method discussed in § 3.2, we construct the $z < 0.06$ bivariate $L_{H\alpha}$ – \mathcal{M} functions and the SMFs of $H\alpha$ SF galaxies. Our results compared to the GAMA SMFs from Baldry et al. (2012) are shown in Figure 15.

Figure 15a left and right panels show the bivariate $L_{H\alpha}$ – \mathcal{M} func-

tions of all SF galaxies (Figure 15a left), and photometrically classified blue and red SF sub-populations (Figure 15a right top and bottom panels, respectively). The bivariate $L_{H\alpha}$ – \mathcal{M} function of all $z < 0.06$ SF galaxies comprise both blue and red galaxies, though dominated by blue ones. The SMFs obtained from integrating the bivariate functions over $L_{H\alpha}$ are shown in Figure 15b, and the GAMA SMFs of Baldry et al. (2012) are also shown for comparison. Baldry et al. (2012) find that the SMF of all $z < 0.06$ GAMA galaxies is well described by a double Schechter function. This functional form has also been used by other authors (e.g. Popesso et al. 2006; Baldry et al. 2008; Drory et al. 2009; Pozzetti et al. 2010) to describe LFs and SMFs, and its origin is related to the bimodal colour–magnitude distribution of (blue and red) galaxies. Pozzetti et al. (2010) and Baldry et al. (2008) find that the massive end of the SMF ($\log \mathcal{M} > 10.5 M_\odot$) is largely dominated by red galaxies while blue galaxies mainly contribute to the faint-end of the SMF. This is also evident in the left panel of Figure 15b where the faint-end of the SMF of all $z < 0.06$ SF galaxies is well matched to that of the GAMA SMF (Baldry et al. 2012), while the bright end of the SMF shows a significant discrepancy. This disagreement arises naturally from our sample consisting only of $H\alpha$ SF galaxies. As only a small fraction of photometrically classified $z < 0.06$ red galaxies have reliably detected $H\alpha$ emission, the GAMA SMF of $z < 0.06$ SF galaxies (left panel of Figure 15b) disagree with the SMF of Baldry et al. (2012) at the high-mass end. In the right top panel of Figure 15, the SMF of blue galaxies (Baldry et al. 2012, open blue squares) and blue SF galaxies (black triangles) are in good agreement for most stellar masses. This indicates that the blue galaxies identified by the colour split advocated by Baldry et al. (2012) are mostly star-forming, with $F_{H\alpha} > 1 \times 10^{-18} \text{ W m}^{-2}$. The SMF of all SF galaxies (grey filled

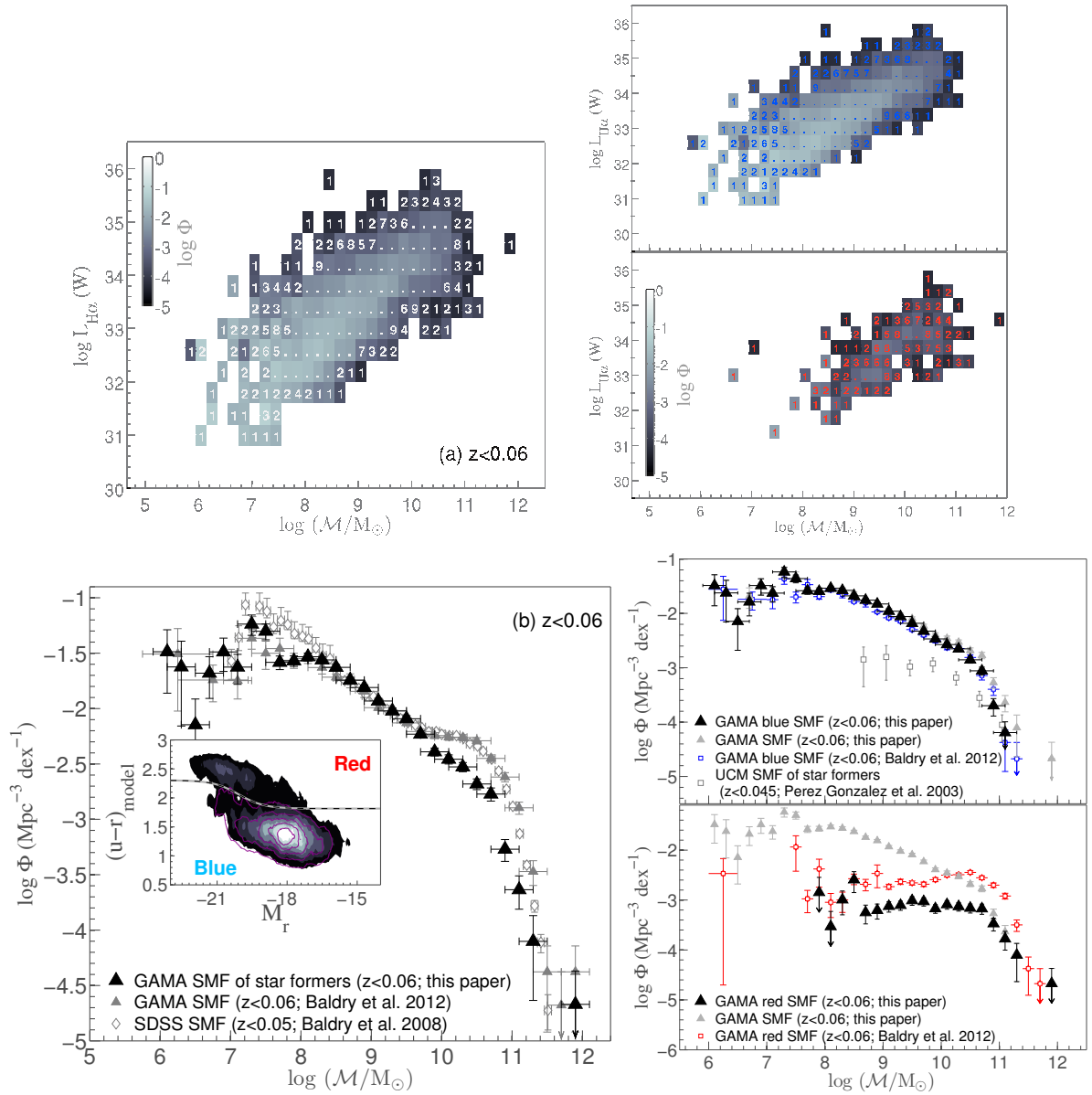


Figure 15. (a) The $z < 0.06$ GAMA bivariate $L_{H\alpha}$ - M functions of all, blue and red SF galaxies (left, top right and bottom right panel respectively), constructed using the density corrected $1/V_{\text{max}}$ method described in § 3.2. The grey scale indicates the log of number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-2}$. (b) The SMFs obtained from integrating the bivariate $L_{H\alpha}$ - M functions over $L_{H\alpha}$ are compared to the $z < 0.06$ GAMA SMFs of Baldry et al. (2012) and $z < 0.05$ SDSS SMF of Baldry et al. (2008).

symbols) shows higher Φ values at higher masses than the GAMA blue SMF as a result of the contribution from the photometrically classified red SF galaxies. The SMF of red SF galaxies is shown in the right bottom panel of Figure 15b. Within uncertainties the shape of the red SF SMF is similar to that of the red SMF indicating that red SF galaxies are a small and stellar mass independent fraction of all red galaxies. Figure 15b (left panel) inset shows the distribution of the $H\alpha$ SF sample (purple contours) compared to all $z < 0.06$ galaxies regardless of star formation, and the dashed line is the colour-magnitude cut used here. Even though photometrically classified blue and red galaxy sub-populations are conventionally labelled as SF and passive galaxies, respectively, a sample selected on the basis of detected $H\alpha$ emission, which is a direct tracer of on-going star formation in a galaxy, includes both photometrically classified blue and red galaxies. Furthermore, not all galaxies with a photometric blue classification have detected $H\alpha$ emission. The fact that some red galaxies are SF while some blue ones are not currently forming stars raises the question: does the

shape of the SMF of blue and red SF galaxies any dependence on the fraction of red star formers and blue non star formers? The evidence of the existence red $H\alpha$ SF galaxies is more pronounced in the colour-magnitude distributions shown in Figure 18, where the number statistics of red SF systems are higher than the $z < 0.06$ sample.

Finally, it is interesting to note that approximately 20–30% of photometrically classified red galaxies contributing to the GAMA red SMF of Baldry et al. (2012) at all stellar masses are star forming. Approximately 40% of the $z < 0.06$ red SF galaxies in our sample are also detected at $250\mu\text{m}$ in Herschel-ATLAS survey, and they cover over two orders of magnitude in $250\mu\text{m}$ luminosity. Given dust is a requirement to be detected at this wavelength, it is reasonable to conclude that at least 40% of the red star formers in our low-redshift sample are dusty SF galaxies and that dust in these galaxies contributes to their redder colour. Moreover, these dusty systems indicate a large spread in $NUV - r$ colour, which is an indicator of recent star formation. This implies that in addi-

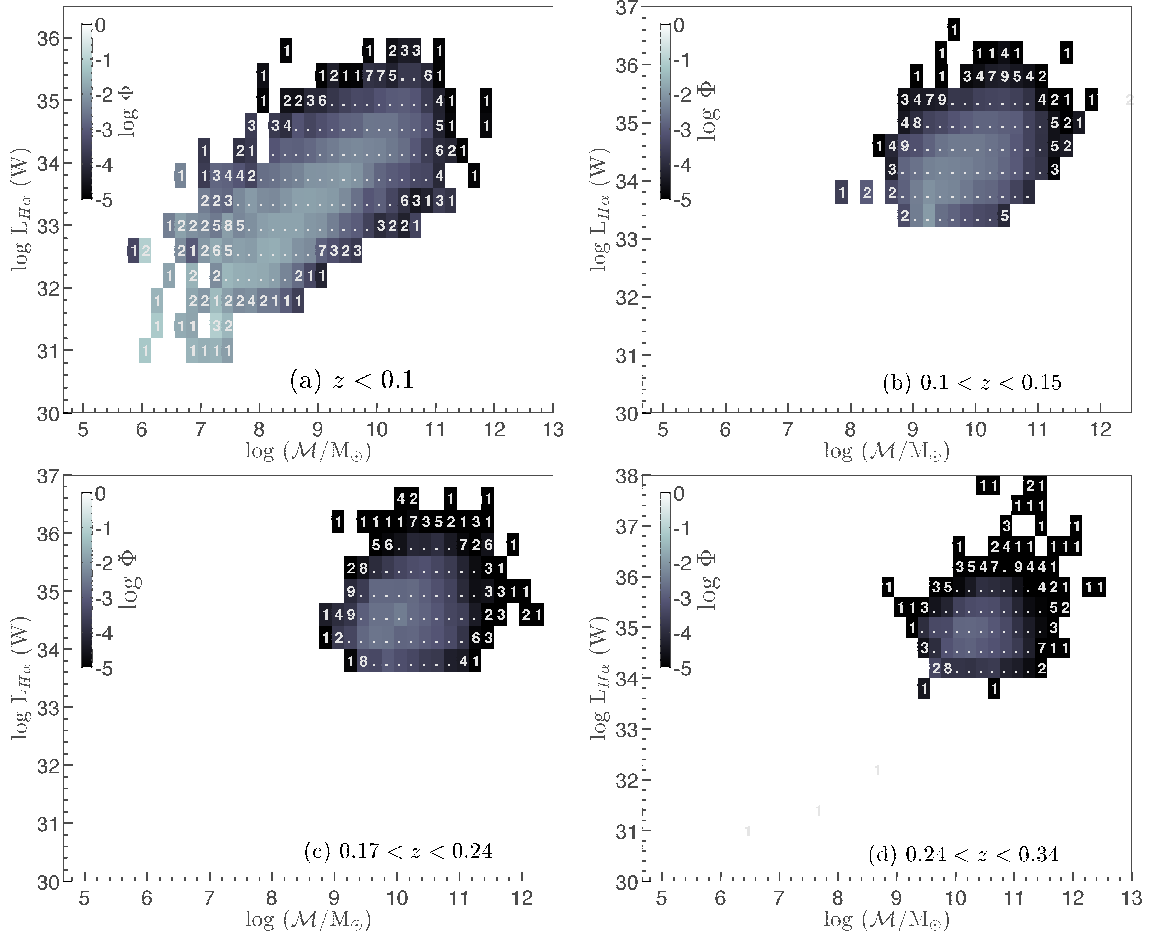


Figure 16. The bivariate $L_{H\alpha}$ – \mathcal{M} functions of all SF galaxies split in to four redshift bins. The grey scale indicates the log number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-2}$.

tion to the dusty on-going star formation these galaxies are also dominated by old stars, and these old stars likely also contribute to the redder galaxy colour.

6.2 The Stellar Mass Function of Star-forming Galaxies Across Time

We divide our sample into four redshift bins to investigate the evolution of the bivariate $L_{H\alpha}$ – \mathcal{M} functions. For each redshift range, we construct the all $H\alpha$ SF (Figure 16), and the respective photometrically classified blue and red SF bivariate functions (Figure 17). The lowest redshift bivariate $L_{H\alpha}$ – \mathcal{M} functions (i.e. all SF, and blue and red SF) probe the largest range in both $L_{H\alpha}$ ($30.5 \lesssim \log L_{H\alpha} \lesssim 36$) and \mathcal{M} ($6 \lesssim \log \frac{\mathcal{M}}{M_{\odot}} \lesssim 12$), while the survey and sample selection effects dominate the higher- z ($0.1 \lesssim z \lesssim 0.34$) bivariate $L_{H\alpha}$ – \mathcal{M} functions as evident from the decrease in $L_{H\alpha}$ and \mathcal{M} ranges.

The SMFs computed from integrating each bivariate function over $L_{H\alpha}$ are compared with the published measurements that cover similar redshift ranges. The results for the $z < 0.1$ range is shown in Figure 18, and 19 show the higher- z ($0.1 \lesssim z \lesssim 0.34$) results. The $z < 0.1$ SMF of all SF galaxies (black filled symbols in all the panels of Figure 18) is mostly in agreement, particularly at the low stellar mass end, with the photometrically classified blue SMFs of Baldry et al. (2004) and Moustakas et al. (2013). The differences seen at higher masses, where the $z < 0.1$ blue SF SMF shows relatively lower amplitudes than other published measurements,

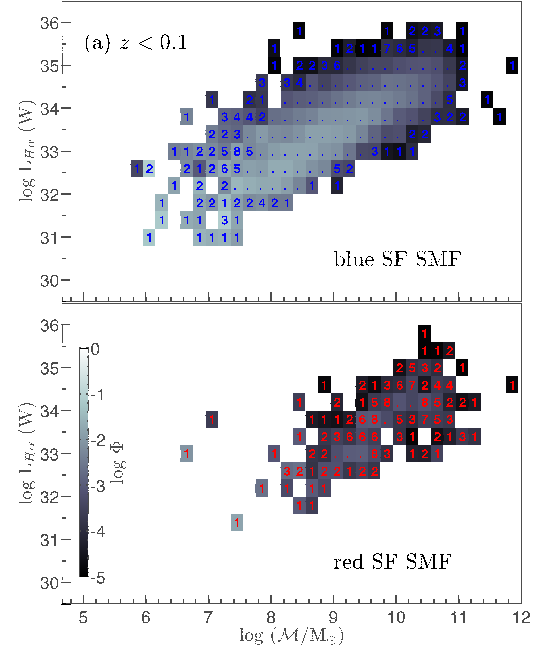


Figure 17. The $z < 0.1$ bivariate $L_{H\alpha}$ – \mathcal{M} functions of blue and red sub-populations. The grey scale indicates the log number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-2}$, and is valid for both blue and red bivariate functions corresponding to the given redshift range.

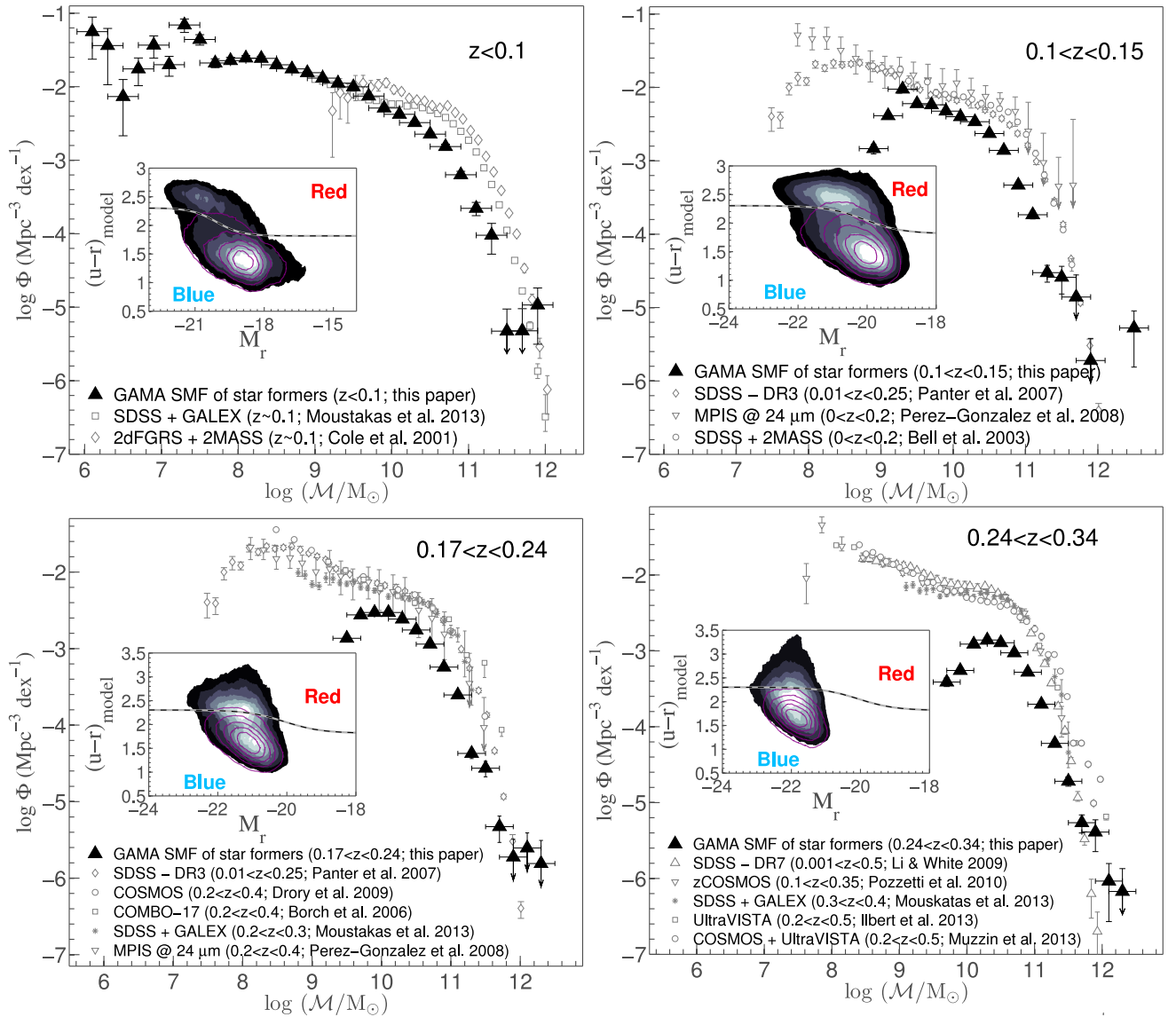


Figure 18. The SMFs derived from integrating the $z > 0.1$ $L_{H\alpha}$ – \mathcal{M} functions shown in Figure 16 over $L_{H\alpha}$ for all SF galaxies. The redshift increases from top-to-bottom. All published SMF measurements shown are adjusted to our assumed cosmology and to a Chabrier (2003) IMF.

highlight that not all photometrically classified blue sources are in fact SF. The same trend can be seen in the higher- z SMFs shown in Figure 19.

Overall, the trends discussed in § 6.1 with regard to the $z < 0.06$ SMF (i.e. the difference between the SMFs of all $H\alpha$ SF galaxies and all galaxies at the high mass end that arises from the lack of many red SF galaxies, and that $\sim 20 - 30\%$ of photometrically classified red galaxies contributing to the red SMFs are likely SF) are also evident in the higher- z SMFs shown in Figures 18 and 19.

6.3 Mass-dependent evolution of the cosmic star formation history

To estimate an integrated stellar mass density (ρ^*), we fit the analytic form introduced in § 5 to the bivariate $L_{H\alpha}$ – \mathcal{M} functions. We assume here that the relationship between $\log L_{H\alpha}$ and M_r derived in § 5 is also valid for $\log L_{H\alpha}$ and $\log \mathcal{M}$, which is not an unreasonable assumption given the observed tight relationship between M_r and $\log \mathcal{M}$. In this interpretation of Eq. 11, $\log \mathcal{M}^0$, the equivalent of M_r^0 , is fixed at $\log \mathcal{M}/M_\odot = 9$. The best fitting

functions for the $z < 0.1$ and $0.1 < z < 0.15$ bivariate $L_{H\alpha}$ – \mathcal{M} functions are shown in the top panels of Figure 20. Our model provides a good description of the observed low redshift bivariate $L_{H\alpha}$ – \mathcal{M} function, especially the faint-end distribution of the function. As the ranges in $L_{H\alpha}$ and \mathcal{M} sampled by the observed $0.1 < z < 0.15$ bivariate $L_{H\alpha}$ – \mathcal{M} function is significantly less than that of the $z < 0.1$ function, the two model parameters, α_L and α_M , that describe the faint-end shape a bivariate function are assumed to be equal to the best-fitting $z < 0.1$ model values for the $0.1 < z < 0.15$ bivariate function. The best-fitting parameters for the $z < 0.1$ and $0.1 < z < 0.15$ redshift bins are given in Table 4.

The relationship between \mathcal{M} and ρ^* for a fixed redshift range is shown in the bottom panel of Figure 20. Overplotted in Figure 20 are the \mathcal{M} and ρ^* relationships based on SDSS data at $z \sim 0.05$ and ROLES data at $z \sim 1$ (Li et al. 2011) and Sobral et al. (2014) HiZELS data at $z \sim 0.4, 0.8, 1.47$ and 2.23 . The GAMA \mathcal{M} and ρ^* relationship agrees well with that of Li et al. (2011) and Sobral et al. (2014). Also, a comparison between GAMA, SDSS, ROLES and HiZELS results indicates that the shape of the relationship does not vary as a function of redshift, rather it is the normalisation

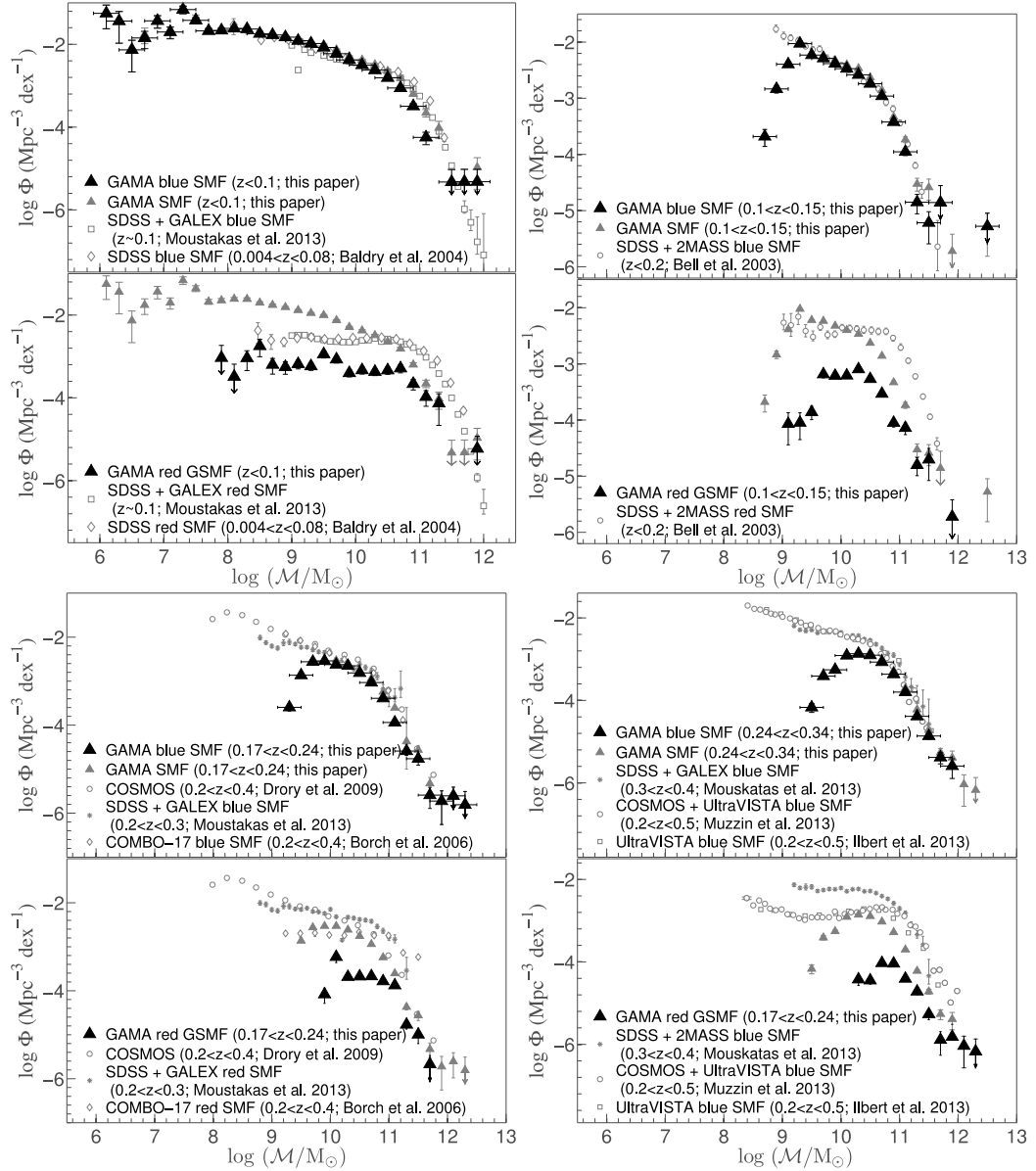


Figure 19. The blue (first panel in each two panel set) and red (second panel in each two panel set) SMFs derived from integrating the blue and red bivariate functions shown in Figures 17 and B1. All published SMF measurements shown are adjusted to our assumed cosmology and to a Chabrier (2003) IMF.

Table 4. The best fitting Schechter–Saunders parameters for the $z < 0.1$ and $0.1 < z < 0.15$ bivariate $L_{H\alpha}$ – \mathcal{M} functions.

Parameter	$z < 0.1$	$0.1 < z < 0.15$
$\log \mathcal{M}^*/M_{\odot}$	10.77 ± 0.07	10.84 ± 0.04
$\alpha_{\mathcal{M}}$	-1.31 ± 0.04	-
$\log \Psi^* (\text{Mpc}^{-3})$	-4.50 ± 1.23	-4.77 ± 0.20
$\log L_{H\alpha}^* (\text{W})$	32.00 ± 0.58	32.06 ± 0.20
β	0.70 ± 0.02	0.67 ± 0.05
σ	0.50 ± 0.02	0.52 ± 0.03
α_L	1.20 ± 0.93	-

of the relationship that change. In the context of galaxy downsizing (Cowie et al. 1999), which states that high-mass galaxies formed their stars early and rapidly while low-mass counterparts formed stars at a slower rate and later times, the peak of the $\log \mathcal{M}$ and ρ^* relationship is expected to shift towards lower masses with decreasing redshift. The lack of such a change contradicts the downsizing scenario, however, as Gilbank et al. (2010) point out high SFR galaxies are also likely to be high-mass systems that both

dominate the galaxy numbers and ρ^* at high z , which could be considered ‘downsizing’.

The ρ^* derived from integrating the best-fitting analytic functions to the bivariate $L_{H\alpha}$ – \mathcal{M} functions (Figure 20) are shown in Figure 21. Also shown for comparison are the ρ^* measurements from recent studies at various redshifts up to $z \sim 3$. Note that all published measurements shown in Figure 21 are calculated from univariate SMFs based on galaxy samples drawn from broadband surveys. The data indicated as lower limits show the cases where the authors have provided a ρ^* measurement by integrating the univariate SMF down to a limiting \mathcal{M} , rather than to zero.

The left panel of Figure 21 compares the GAMA ρ^* measurements based on bivariate $L_{H\alpha}$ – \mathcal{M} functions with those derived from total galaxy samples, and the right panel of the same figure compares our results with the ρ^* based on photometrically classified blue galaxy populations and emission line samples. Our results agree well with the published measurements based on photometrically classified blue galaxy populations and emission line sam-

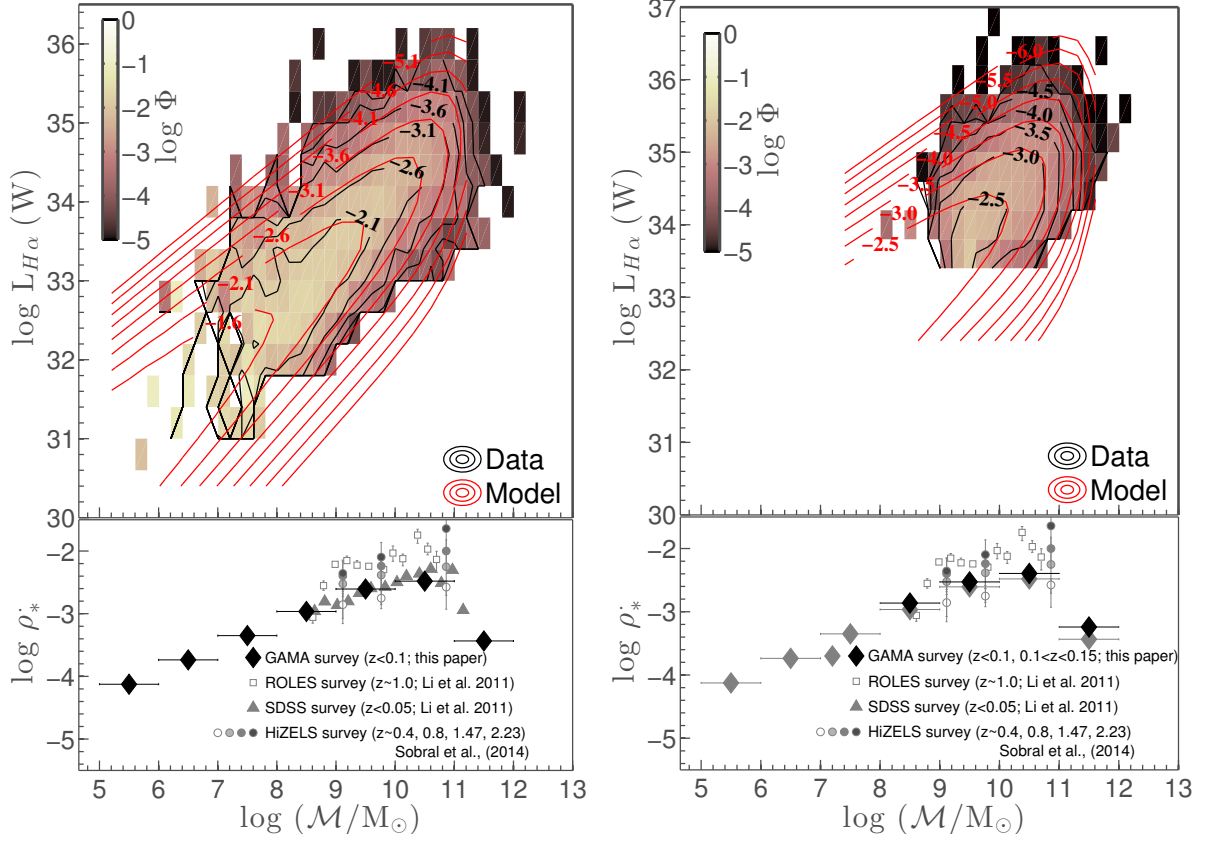


Figure 20. Top panels: Best-fitting analytic models (red lines) to the $z < 0.1$ (left) and $0.1 < z < 0.15$ (right) bivariate $L_{H\alpha}$ - M functions. The colour scale and black contours indicate the log number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-2}$. Bottom panels: SFRD ($\text{M}_{\odot} \text{yr}^{-1} \text{Mpc}^{-3} \text{dex}^{-1}$) versus stellar mass for the $z < 0.1$ (left) and $0.1 < z < 0.15$ (right). For reference, we show the Li et al. (2011) results based on SDSS (filled triangles) and ROLIS (open squares) surveys and the Sobral et al. (2014) results based on HiZELS (circles) survey. All measurements are adjusted to the Baldry & Glazebrook (2003) IMF.

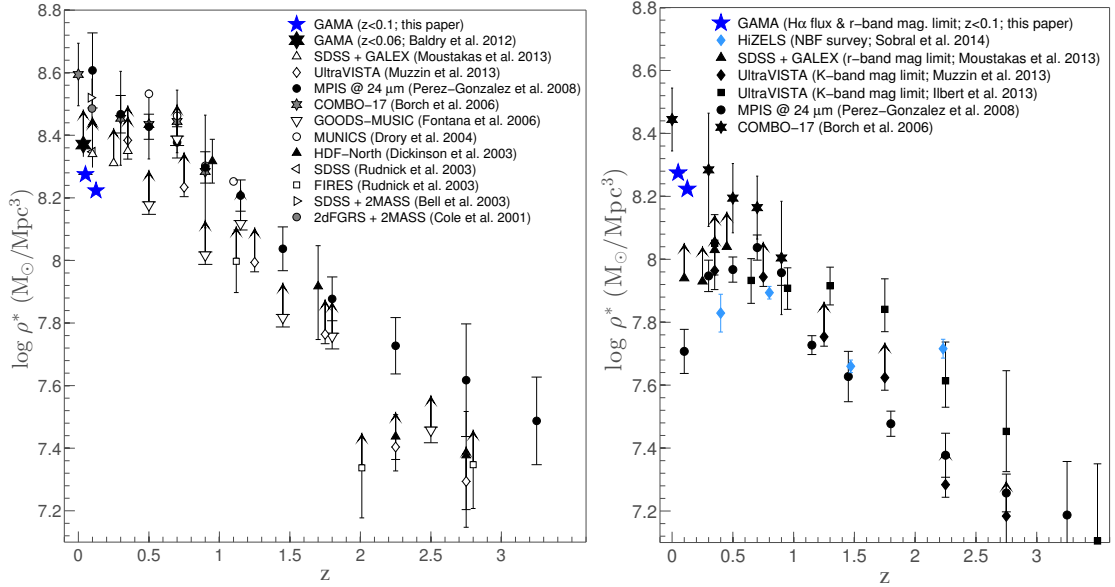


Figure 21. The evolution of the ρ^* , as traced by all galaxies (left) and blue/emission line galaxies (right). GAMA measurements are derived from integrating the analytic forms shown in the top panels of Figure 20. In the right panels photometrically classified blue galaxies are shown with black symbols, while coloured symbols refer to emission line samples. The cases where the authors report a density value evaluated assuming a lower mass limit are shown as lower limits. All measurements are adjusted for our assumed cosmology and for a Chabrier (2003) IMF.

ples, as might be expected given the close agreement between our SF SMFs and the blue galaxy SMF of Baldry et al. (2012).

7 DISCUSSION

In this followup paper to paper I, we have explored the bivariate $L_{H\alpha}$ - M_r and $L_{H\alpha}$ - \mathcal{M} functions of GAMA galaxies. One of the main aims of this analysis is to investigate whether we can reliably recover a correction for the incompleteness introduced in selecting $H\alpha$ detected objects from a r -band limited survey by modelling the low redshift bivariate function, which can then be used as a reference to account for the missing optically faint SF galaxies at higher- z ($0.1 \lesssim z \lesssim 0.34$). Other goals of this investigation include exploring the evolution of bivariate and univariate functions of all $H\alpha$ SF and photometrically classified blue and red $H\alpha$ SF sub-populations relative to the evolution of all, blue and red populations regardless of star formation, and the mass dependence of the SFR history.

In order to compare our results more directly with the previous GAMA LF (Loveday et al. 2012) and SMF (Baldry et al. 2012) studies, we adopt the LF estimators used in their studies to construct the bivariate functions. A discussion on the formulation of three LF estimators (e.g. the classical method, density corrected $1/V_{\max}$ method and SWML estimator) for constructing bivariate functions is presented in § 3 and in Appendix A. A comparison between the bivariate functions based on these methods (§ 3.3) shows that the differences in number densities are largely limited to the faint $L_{H\alpha}$ - M_r end of the bivariate functions as expected given the relatively small volumes sampled. The bivariate $L_{H\alpha}$ - M_r functions and LFs presented in § 4 and $L_{H\alpha}$ - \mathcal{M} functions and SMFs presented in § 6.2 are based on the classical $1/V_{\max}$ method, and are compared with the GAMA LFs of Loveday et al. (2012) and other published measurements that are mostly based on the classical method. The bivariate $L_{H\alpha}$ - \mathcal{M} functions and SMFs presented in § 6.1 are based on the density corrected $1/V_{\max}$, and are compared with the GAMA SMFs (Baldry et al. 2012) based on the same method.

As a consequence of the magnitude limited nature of the GAMA survey, the bivariate $L_{H\alpha}$ - M_r functions (Figure 6 and 7) show a progressive decline in number density (Φ) towards fainter $L_{H\alpha}$ and M_r , and the range in $L_{H\alpha}$ and M_r probed decrease with increasing redshift. In each of the four redshift bins considered, the $L_{H\alpha}$ - M_r range, and the Φ values of the blue $H\alpha$ SF bivariate functions (Figure 7) are similar to that of the total $H\alpha$ SF functions (Figure 6). This result indicates that the galaxies contributing to the total $H\alpha$ SF bivariate functions are mostly drawn from the photometrically classified blue sub-populations at each redshift. This is further corroborated by both the relatively smaller range in $L_{H\alpha}$ and M_r probed by the red bivariate $L_{H\alpha}$ - M_r functions (Figure B2), and the close agreement between the Loveday et al. (2012) blue LFs and those computed by integrating the blue bivariate $L_{H\alpha}$ - M_r functions at all redshifts (Figures 8 and B2). While the red $H\alpha$ SF bivariate number density contribution to the total $H\alpha$ SF bivariate function is relatively low, Figures 8 and B2 show that a fraction of galaxies classified as red at all M_r values are in fact SF galaxies. The same conclusion is drawn the bivariate $L_{H\alpha}$ - \mathcal{M} functions (Figures 15, 16 and 17), and the SMFs (Figures 18, 19) computed from integrating the bivariate functions. While the $H\alpha$ SF galaxy population at each redshift is primarily drawn from the blue sub-populations at that redshift, approximately 20 – 30% of those photometrically classified as red and conventionally called passive are forming massive stars at all stel-

lar masses at each redshift range probed. We find that $\sim 40\%$ of the red $H\alpha$ SF galaxies at $z < 0.06$ (i.e. those contributing to the red bivariate $L_{H\alpha}$ - \mathcal{M} function and the SMF shown in Figure 15) is also reliably detected at $250\mu m$. As dust is a requirement to be able to be detected at this wavelength, it is likely that some of the red $H\alpha$ SF galaxies are dusty (and therefore red) SF systems. Moreover, those detected at $250\mu m$ cover a large range in $NUV - r$ colour, which is an indicator of recent star formation in galaxies, and most lie below the $NUV - r = 5.4$ cutoff for recent star formation from Schawinski et al. (2007). This points to red $H\alpha$ SF galaxies having a non-negligible fraction of underlying old stellar population that likely also contribute to their redder colour.

Motivated by the analytic formalism widely used to model bivariate brightness profiles (e.g. de Jong & Lacey 2000; Cross et al. 2001; Driver et al. 2005; Ball et al. 2006a), in § 5.1 we introduce a simple analytic model to describe the observed bivariate functions presented in § 4 and 6. This model assumes that the bivariate function can be written as a product of two functions (Choloniewski 1985; Corbelli et al. 1991), a Schechter (1976) function representative of r -band LFs (or SMFs) and a Saunders et al. (1990) function representative of $H\alpha$ LFs (paper I). The two functional forms are linked in the bivariate analytic relation through the observed relationship between $L_{H\alpha}^*$ and M_r (Figure 11). The resultant best-fitting bivariate models to the $z < 0.1$ $L_{H\alpha}$ - M_r , and the $z < 0.1$ and $0.1 < z < 0.15$ $L_{H\alpha}$ - \mathcal{M} functions (Figures 12 and 20, respectively) provide a good description of the data, and from integrating the best-fitting $z < 0.1$ bivariate $L_{H\alpha}$ - M_r model we were able to recover the $z < 0.1$ SFRD reported in paper I. The same model is used to fit the rest of the higher- z bivariate functions by fixing the faint $L_{H\alpha}$ - M_r (or \mathcal{M}) end slope of the bivariate model to be equal to that of the lowest redshift best-fitting model. As the $0.1 < z < 0.15$ bivariate functions sample a relatively large range in both $L_{H\alpha}$ and M_r (or \mathcal{M}), the $0.1 < z < 0.15$ models can be reasonably well constrained, however, the two higher- z models cannot be constrained accurately even with assuming a constant faint-end slope (i.e. assuming no evolution in the faint-end slope of the bivariate function) for the model. This is mainly due to the observed higher- z bivariate $L_{H\alpha}$ - M_r and $L_{H\alpha}$ - \mathcal{M} functions being incomplete at around the characteristic $L_{H\alpha}$, M_r and \mathcal{M} values. By integrating the best-fitting higher- z bivariate $L_{H\alpha}$ - M_r and $L_{H\alpha}$ - \mathcal{M} models, particularly the $0.1 < z < 0.15$ models which we were able to constrain more accurately than the rest, we were able to recover an approximate correction for the missing optically faint SF galaxies. The SFRDs estimated this way (Figure 14) show a strong evolution with redshift up to $z \sim 0.2$ and a flattening thereafter, though, we caution against using the last two points (i.e. the two higher- z ρ^* measurements) in understanding the evolution of ρ^* with redshift as their best-fitting bivariate models are not well constrained.

The mass dependence of the SFR history and the evolution of the stellar mass density of $H\alpha$ SF galaxies are explored in § 6.3. The \mathcal{M} and ρ^* relationship for the $z < 0.1$ and $0.1 < z < 0.15$ redshift ranges derived from integrating the respective GAMA bivariate $L_{H\alpha}$ - \mathcal{M} functions is shown in Figure 20. Also shown for comparison are the \mathcal{M} - ρ^* relations derived using SDSS data at $z \sim 0.05$, ROLES data at $z \sim 1$ (Li et al. 2011) and HiZELS data at $z \sim 0.4, 0.8, 1.47, 2.23$ (Sobral et al. 2014). A comparison between GAMA, SDSS, ROLES and HiZELS \mathcal{M} - ρ^* results show that while the normalisation of the \mathcal{M} - ρ^* relation increases modestly with redshift, the shape remains the same. This result points towards a scenario where SFRs in all galaxies, regardless of stellar mass, decline with decreasing redshift (Karim et al. 2011; Sobral et al. 2014), contradicting the ‘galaxy downsizing’ scenario of Cowie et al. (1999), which states that massive galaxies formed

their stars early at a rapid rate while the low-mass systems formed stars late at a slower rate. However, as Gilbank et al. (2010) point out the high SFR systems are also likely to be high-mass galaxies that dominate SFRD at high- z , which could be considered as downsizing.

Finally, the evolution of the ρ^* of H α SF population in comparison to the evolution of ρ^* of all galaxies regardless of star formation, and of photometrically classified blue galaxies (black markers) and other H α galaxy samples (coloured markers) is shown in the left and right panels of Figure 21, respectively. The SMDs based on the H α SF sub sample of GAMA galaxies lie low in the left panel of Figure 21. This is expected as other stellar mass densities shown in that panel are based on all galaxies regardless of star formation. Our results are in good agreement with the SMDs based only on either the blue sub-population of galaxies or other emission-line samples (right panel of Figure 21). The scatter in data points is most likely due to cosmic (sample) variance.

8 SUMMARY

We have explored the GAMA bivariate $L_{H\alpha}$ - M_r and $L_{H\alpha}$ - \mathcal{M} functions in the paper, and the key results of the analysis are as follows:

- i. By modelling the low redshift distribution of the bivariate $L_{H\alpha}$ - M_r function, we estimate a correction for the missing optically faint star forming galaxies at higher- z . The corrected stellar mass densities presented in Figure 14 show high level of consistency with earlier published results (e.g. Hopkins & Beacom 2006), suggesting that this approach is reasonable. The implication is that the shape of the faint-end of the bivariate functions does not evolve strongly over this redshift range.
- ii. The H α SF sample used for this study consists of both photometrically classified blue and red galaxies, though dominated by blue galaxies. This allows us to construct not only the bivariate and univariate LFs and SMFs of H α SF galaxies, but also the LFs and SMFs of photometrically classified blue and red SF galaxies.
- iii. The M_r LFs obtained from integrating the bivariate $L_{H\alpha}$ - M_r functions are in agreement with the GAMA M_r LFs of photometrically classified blue galaxies from Loveday et al. (2012). Also, the $z < 0.06$ SMF of SF galaxies obtained from integrating the bivariate $L_{H\alpha}$ - \mathcal{M} function is consistent with the blue SMF of Baldry et al. (2012).
- iv. The low redshift ($z < 0.15$) ρ^* and \mathcal{M} relationship derived using GAMA data agrees well the $z < 0.05$ results from Li et al. (2011). The shape of the ρ^* and \mathcal{M} relationship at low redshift is similar to that obtained at $z \sim 1$ from Gilbank et al. (2010). The comparison of results between different redshifts indicate that the shape of the ρ^* and \mathcal{M} relationship does not change with redshift over the redshift range probed by the bivariate functions, rather it is the normalisation of the relationship at all masses that varies with redshift.
- v. The GAMA stellar mass densities based on bivariate $L_{H\alpha}$ - \mathcal{M} functions, i.e. based on a sample of H α SF galaxies, is in good agreement with the published stellar mass densities based on photometrically classified blue galaxy populations.

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GAMA is a joint European-Australasian project based around a spectroscopic campaign using the Anglo-Australian Telescope. The GAMA input catalogue is based on data taken from the Sloan Digital Sky Survey and the UKIRT Infrared Deep Sky Survey. Complementary imaging of the GAMA regions is being obtained by a number of independent survey programs including GALEX MIS, VST KIDS, VISTA VIKING, WISE, Herschel-ATLAS, GMRT and ASKAP providing UV to radio coverage. GAMA is funded by the STFC (UK), the ARC (Australia), the AAO, and the participating institutions. The GAMA website is <http://www.gama-survey.org/>.

Data used in this paper will be available through the GAMA website (<http://www.gama-survey.org/>) once the associated redshifts are publicly released.

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APPENDIX A: BIVARIATE STEPWISE MAXIMUM LIKELIHOOD METHOD

In order to overcome the inconvenience of not being able to adequately visualise whether the chosen parameterisation represents a good fit to the data (Willmer 1997), a “nonparametric” LF stepwise maximum likelihood (SWML) LF estimator is introduced by Efstathiou et al. (1988). The SWML method does not suffer from the same biases that affect $1/V_{\text{max}}$ technique. Unlike $1/V_{\text{max}}$ method, the luminosity bins in SWML are highly correlated, so any issue that occurs affect the whole luminosity function.

Efstathiou et al. (1988) describes the SWML formulation for univariate LFs. An extension of the SWML LF estimator for bivariate LFs is discussed in Sodre & Lahav (1993) for the bivariate diameter–luminosity function and in Driver et al. (2005) for the bivariate brightness distribution. Following Efstathiou et al. (1988); Sodre & Lahav (1993) and Driver et al. (2005), we construct a bivariate SWML estimator for $H\alpha/M_r$ and $H\alpha/\text{stellar mass}$ LFs.

The probability of observing a galaxy i with a $H\alpha$ luminosity L_i and an absolute r -band magnitude M_{r_i} at a redshift z_i , inclusive of the completeness/selection function (f), is defined to be (Sodre

& Lahav 1993; Loveday 2000; Ball et al. 2006a)

$$p_i \propto \frac{\Phi(M_i, L_i) f(M_i, L_i, z_i)}{\int_{L_{faint}(z_i)}^{L_{bright}(z_i)=\infty} \int_{M_{faint}(z_i)}^{M_{bright}(z_i)} \Phi(M, L) f(M, L, z) dM dL} \quad (\text{A1})$$

The probability, p_i , is directly proportional to the differential LF at M_i and L_i , and inversely proportional to the faintest and brightest absolute magnitudes (M_{faint} , M_{bright}) and $H\alpha$ luminosities (L_{faint} , L_{bright}) visible at z_i (Willmer 1997; Heyl et al. 1997). Note that we do not account for survey incompleteness via f . Instead the incompleteness corrections defined in paper I are incorporated into the bivariate LF through a weighting function (Driver et al. 2005) as explained below.

The SWML LF is derived by maximising the likelihood function, $\mathcal{L} = \prod_{i=1}^{N_g} p_i$, generally $\log \mathcal{L}$, with respect to the discretised luminosity distribution. The bivariate LF, $\Phi(M, L)$, is parameterised as N_M and N_L steps (Efstathiou et al. 1988):

$$\Phi(M, L) = \Phi_{jk} \quad j=1 \dots N_M, \text{ and } k=1 \dots N_L \quad (\text{A2})$$

where N_M and N_L are evenly spaced bins in M_r and $H\alpha$ luminosity with $M_j - \frac{\Delta M}{2} \leq M_i \leq M_j + \frac{\Delta M}{2}$, and $L_k - \frac{\Delta L}{2} \leq L_i \leq L_k + \frac{\Delta L}{2}$.

Rewriting the denominator of Eq. A1 in summation notation gives the following log-likelihood function

$$\ln \mathcal{L} = \sum_{i=1}^{N_g} \sum_{j=1}^{N_M} \sum_{k=1}^{N_L} W[M_i - M_j, L_i - L_k] \ln \Phi_{jk} - \sum_{i=1}^{N_g} \ln \left\{ \sum_{a=1}^{N_M} \sum_{b=1}^{N_L} \Phi_{ab} \Delta M \Delta L H[M_a - M_{faint}(z_i), L_b - L_{faint}(z_i)] \right\} + C, \quad (\text{A3})$$

where C is a constant, $W[M_i - M_j, L_i - L_k]$ is the weighting function defined as (Driver et al. 2005)

$$W[M_i - M_j, L_i - L_k] = \begin{cases} 1 & \text{if } M_j - \frac{\Delta M}{2} \leq M_i \leq M_j + \frac{\Delta M}{2} \\ & \text{and } L_k - \frac{\Delta L}{2} \leq L_i \leq L_k + \frac{\Delta L}{2} \\ 0 & \text{otherwise,} \end{cases} \quad (\text{A4})$$

and $H[M_j - M_{faint}(z_i), L_k - L_{faint}(z_i)]$ is the ramp function, inclusive of incompleteness corrections (c; paper I), defined as,

$$H[M_j - M_{faint}(z_i), L_k - L_{faint}(z_i)] = \frac{1}{\Delta M \Delta L} \int_{M_j - \frac{\Delta M}{2}}^{M'} dM \int_{L'}^{L_k + \frac{\Delta L}{2}} dL O_i(M, L) c. \quad (\text{A5})$$

The definitions of M' and L' are:

$$\begin{aligned} M' &= \max\{M_j - \frac{\Delta M}{2}, \min[M_j + \frac{\Delta M}{2}, M_{faint}(z_i)]\} \\ L' &= \max\{L_k - \frac{\Delta L}{2}, \min[L_k + \frac{\Delta L}{2}, L_{faint}(z_i)]\} \end{aligned}$$

Simply setting M' or L' to be equal to the faint magnitude or luminosity bin boundary will result in the LF being underestimated in incompletely sampled bins. If the faint bin boundaries are used in the ramp function, then the incomplete bins should be excluded (Loveday et al. 2012). Finally, the function, $O_i(M, L)$, in Eq. A5 is the observable window function for each galaxy at z_i (Driver et al. 2005), and has the following form.

$$O_i(M, L) = \begin{cases} 1 & \text{if } M_{bright,i} \leq M_i \leq M_{faint,i} \\ & \text{and } L_{faint,i} \leq L_i \\ 0 & \text{otherwise,} \end{cases} \quad (\text{A6})$$

where,

$$\begin{aligned} M_{bright,i} &= m_{bright} - 5 \log d_{L(z_i)} - 25 - K(z_i) \\ M_{faint,i} &= m_{faint} - 5 \log d_{L(z_i)} - 25 - K(z_i) \\ L_{faint,i} &= 4\pi d_{L(z_i)}^2 f_{H\alpha, faint} \end{aligned}$$

and

$$\begin{aligned} m_{bright} &= 14.65 \quad (\text{SDSS bright magnitude limit}) \\ m_{faint} &= 19.4 \quad \text{or } 19.8 \quad (\text{GAMA faint magnitude limits}) \\ f_{faint} &= 1 \times 10^{-18} \text{ W/m}^2 \quad (H\alpha \text{ flux limit; paper I}) \end{aligned}$$

As discussed in Efstathiou et al. (1988) a constraint must be imposed on the likelihood to fix the normalization constant in Eq. A3, by using a Lagrangian multiplier (λ). We adopt the constraint used by Ball et al. (2006b) and Sodre & Lahav (1993).

$$g = \sum_{j=1}^{N_M} \sum_{k=1}^{N_L} \Phi_{jk} \Delta M \Delta L - 1 = 0. \quad (\text{A7})$$

The likelihood with the constraint applied, $\ln \mathcal{L}' = \ln \mathcal{L} + \lambda g(\Phi_{jk})$, is maximised with respect to Φ_{jk} and λ , requiring $\lambda = 0$. The constraint, although it does not affect the shape of the LF determined by Φ_{jk} , plays a role in the error determination (Efstathiou et al. 1988). The maximum likelihood (i.e. $\partial \ln \mathcal{L}' / \partial \Phi_{jk} = 0$) is then given by,

$$\Phi_{jk} \Delta M \Delta L = \frac{\sum_{i=1}^{N_g} W[M_i - M_j, L_i - L_k]}{\sum_{i=1}^{N_g} \left\{ \frac{H[M_j - M_{faint}(z_i), L_k - L_{faint}(z_i)]}{\sum_{a=1}^{N_M} \sum_{b=1}^{N_L} \Phi_{ab} \Delta M \Delta L H[M_a - M_{faint}(z_i), L_b - L_{faint}(z_i)]} \right\}}, \quad (\text{A8})$$

where, Φ_{ab} is from the previous iteration.

The bivariate SWML LF (Φ_{jk}), by construction, loses the information regarding the absolute normalisation (Efstathiou et al. 1988). We achieve a normalisation for the bivariate SWML LF results presented in this paper by matching the bright-end of the bivariate SWML LFs to their $1/V_{\max}$ LFs. While the shape of the faint-end $1/V_{\max}$ LF can be affected by the over/under densities, this estimator provides reliable abundances at higher luminosities, where it probes a relatively larger volume (Driver et al. 2005; Eke et al. 2005). Therefore, matching to the bright end of the LFs is a robust approach to fix the normalisation of the bivariate SWML LF.

The LF errors can be determined using the fact that maximum likelihood estimates (Φ_{jk}) are asymptotically normally distributed with the covariance matrix.

$$\text{cov}(\Phi_{jk}) = \mathbf{I}^{-1}(\Phi_{jk}), \quad (\text{A9})$$

where \mathbf{I} is the information matrix,

$$\mathbf{I}(\Phi_{jk}) = - \begin{pmatrix} \frac{\partial^2 \ln \mathcal{L}}{\partial \Phi_{ik} \partial \Phi_{jk}} + \frac{\partial g}{\partial \Phi_{ik}} \frac{\partial g}{\partial \Phi_{jk}} & \frac{\partial g}{\partial \Phi_{jk}} \\ \frac{\partial g}{\partial \Phi_{ik}} & 0 \end{pmatrix}.$$

The $H\alpha$ and M_r LFs can be recovered from Eq. A8 by summing over M_r and $L_{H\alpha}$ respectively. For example, to recover the $H\alpha$ univariate LF,

$$\phi_{H\alpha} = \sum_{j=1}^{N_M} \phi_{jk} \times \Delta M. \quad (\text{A10})$$

Figure A1 shows the residual between the bivariate LFs based on the density corrected V_{\max} and bivariate SWML methods. Again the differences are primarily in the faintest bins of the lowest redshift sample. The differences here are greater than seen with the density corrected $1/V_{\max}$ method. Given the small number of

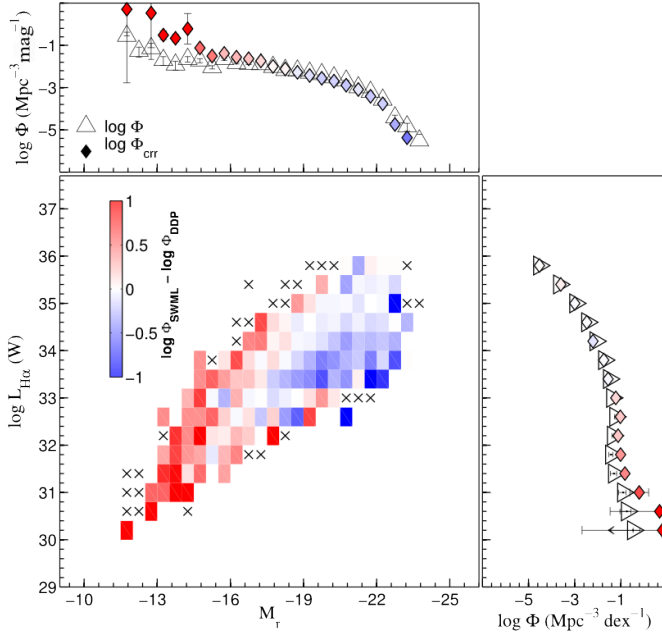


Figure A1. The residual map showing the difference in number densities calculated using the density corrected $1/V_{\text{vmax}}$ ($\log \Phi_{DDP}$) and SWML ($\log \Phi_{SWML}$) methods for the $z < 0.1$ bivariate functions. The crosses indicate where the galaxy number statistics are low. Top and right panels show the univariate LFs inferred from the bivariate functions.

galaxies contributing to these faintest bins, the discrepancy here should be interpreted primarily as a limit on the systematic uncertainty of the measurement of the faint-end of the bivariate function, from all of these methods.

APPENDIX B: BIVARIATE AND UNIVARIATE FUNCTIONS

The higher- z bivariate $L_{H\alpha}$ - \mathcal{M} functions and the univariate LF are shown in Figures B1 and B2.

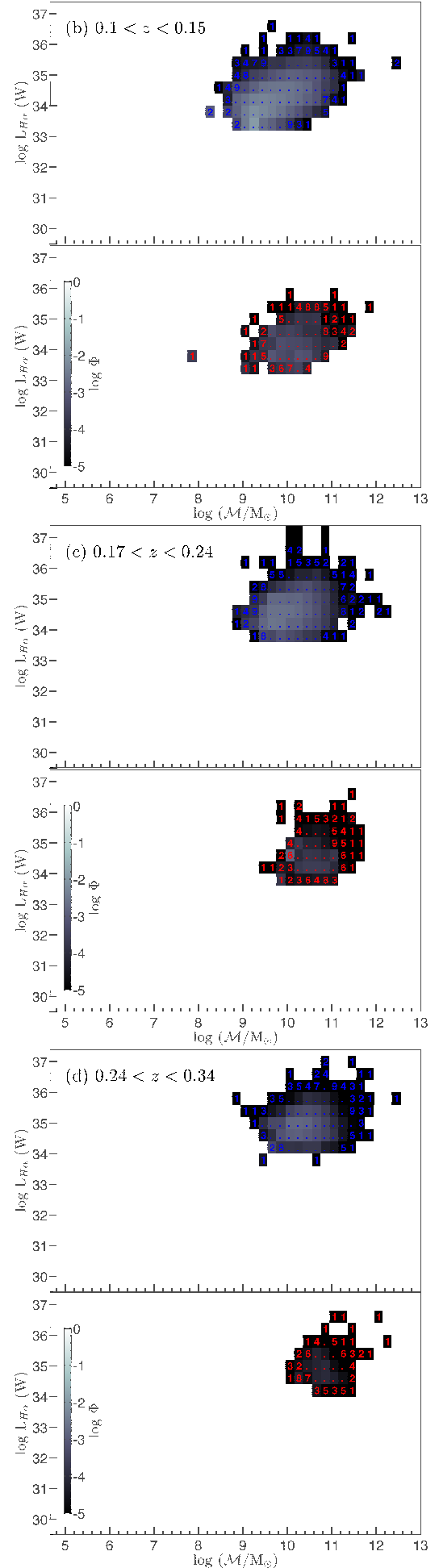


Figure B1. The bivariate $L_{H\alpha}$ - \mathcal{M} functions of blue and red sub-populations split into four redshift ranges. The grey scale indicates the log number densities (Φ) in the unit of $\text{Mpc}^{-3} \text{dex}^{-2}$, and is valid for both blue and red bivariate functions corresponding to the given redshift range.

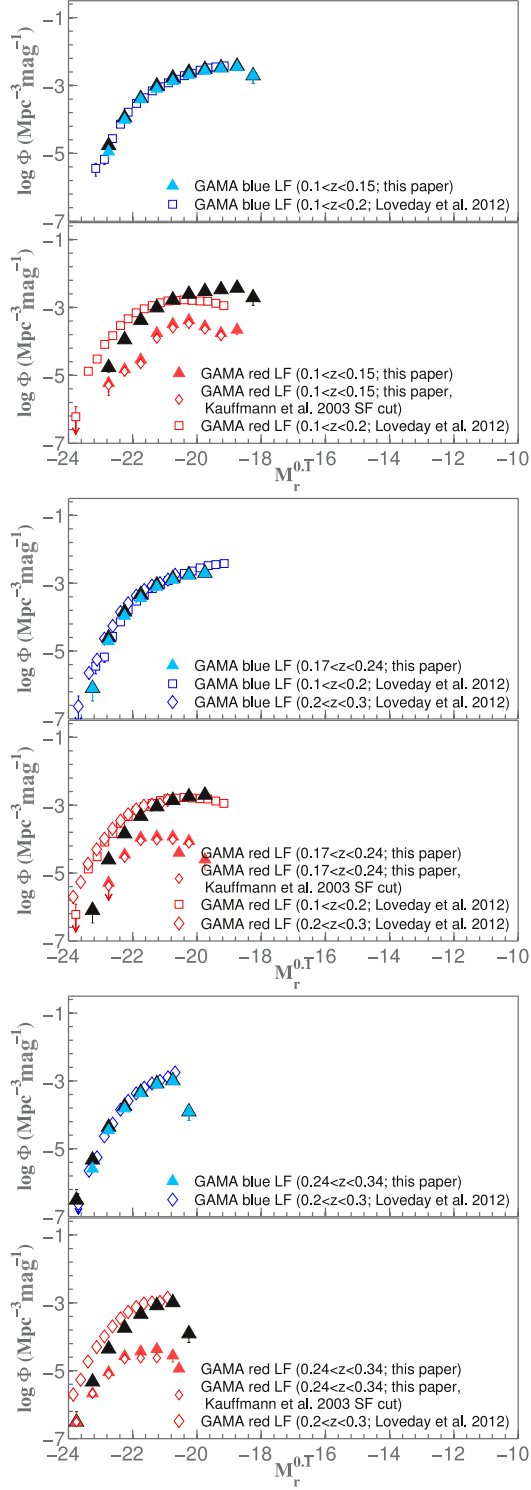


Figure B2. Each two sets of panels from top to bottom show the blue and red SF M_r functions derived from integrating the bivariate $L_{H\alpha}$ - M_r functions shown in Figure 7 over $L_{H\alpha}$. The key indicate the redshift ranges probed. The all SF M_r functions are shown in Figures 8 and 9. The colour cut given in Eq. 4 used to classify galaxies as blue and red. The open red diamonds in each red SF M_r functions panels indicate the red SF M_r function constructed using the Kauffmann et al. (2003) SF/AGN prescription instead of that of Kewley & Dopita (2002).