Predicting the Radio Continuum - Galaxy Emission in SHARK, the Semi-Analytical Model of Galaxy Formation

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Declaration

This is to certify that:

- 1. This thesis comprises of my original work.
- 2. Due acknowledgement has been made in the text to all other materials used.
- 3. This thesis consists of no more than 40 pages inclusive of tables and figures but exclusive of references and appendices.

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I would like to acknowledge that this work has been completed at the University of Western Australia, which is situated on Noongar land, and that the Noongar people remain the spiritual and cultural custodians of this land.

Abstract

TEXT

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Chapter 1

Introduction

1.1 Introduction

1.1.1 The Far-Infrared Radio Correlation

Planck Collaboration et al. (2016) The Far-Infrared Radio Correlation (FIRC) is an observed tight correlation between a population of galaxy's total Far-Infrared (FIR) luminosity and total radio luminosity that spans five orders of magnitude (Van der Kruit (1971), Van Der Kruit (1973), De Jong et al. (1985), Helou et al. (1985), Condon (1992)). It has been shown to exists in a variety of different galactic populations including Sub-Millimetre Galaxies (SMGs) (Thomson et al. (2014), Algera et al. (2020)), (Ultra)-Luminous Infrared Galaxies (ULIRGs) (Lo Faro et al. (2015)), Early Type galaxies (Omar & Paswan (2018)), dwarf galaxies (Shao et al. (2018)), low-ionization nuclear emission-line region (LINERs) and Seyferts (Solarz et al. (2019)), irregular and disk-dominated galaxies (Pavlović (2021)) as well as highly lensed galaxies (Giulietti et al. (2022)), to name a few.

It is often parameterised as it first was in Helou et al. (1985) as in equation ??

$$q_{ir} = \log_{10} \left(\frac{L_{IR}}{3.75 \times 10^{12} \text{Wm}^{-2}} \right) - \log_{10} \left(\frac{L_{1.4 \text{GHz}}}{\text{Wm}^{-2} \text{Hz}^{-1}} \right), \tag{1.1}$$

where L_{IR} is total FIR integrated over $8-1000\mu m$ in the rest frame and $L_{1.4GHz}$ the total rest frame radio luminosity at 1.4GHz. While this is the most common parameterisation (Helou et al. (1985), Condon (1992), Bell (2003) etc.) the FIRC has been shown to exist using different frequency combinations as well (ie radio luminosity at 150MHz (Read et al. (2018)), infrared being a combination of $60\mu m$ and $80\mu m$ (Yun et al. (2001)) and a variety of different individual infrared frequencies (Smith et al. (2014)).

The FIRC's persistence across different magnitudes of luminosity and varied galactic populations makes it an attractive tool. It's uses include identifying Active Galactic Nuclei (AGNs) (Donley et al. (2005), Norris et al. (2006), Algera et al. (2020)), in defining the SFR-Radio correlation (Bell (2003), Condon & Ransom (2016), Duncan et al. (2020), Molnár et al. (2021)), in determining the distance to SMGs (Yun & Carilli (1999)) and deriving a galaxy's radio emission based on

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its IR emission (Dale et al. (2014)).

It is therefore important that we totally understand the FIRC and its limitations. In particular understanding any possible evolution the FIRC has with redshift (z). This is a very active area of research into FIRC and has some studies finding that there is some evolution with z (Ivison et al. 2010a,b; Magnelli et al. 2015; Delhaize et al. 2017) and others finding no evolution (Appleton et al. 2004; Jarvis et al. 2010; Sargent et al. 2010a,b; Bourne et al. 2011; Mao et al. 2011; Duncan et al. 2020; Thomson et al. 2014; Algera et al. 2020). Delvecchio et al. (2021) did not find that qir evolved significantly with z, but instead evolved with stellar mass. Reasons that have been suggested for these differences include choice of spectral index (Delhaize et al. (2017)), use of IR or radio biased samples (Sargent et al. (2010b)) and use of statistically insignificant samples (Jarvis et al. (2010)). However the most commonly cited reason is AGN contamination. Even high precision instruments are biased towards high mass and low redshift objects and determining the presence of an AGN in high redshift galaxies is particularly difficult (Delvecchio et al. (2021)).

Despite the wealth of research that has been conducted into the FIRC, our theoretical understanding is quite limited. Fundamentally the FIRC is grounded in star-formation where galaxy's IR and Radio emission come from the same young, massive stars. These stars, formed in dusty, molecular clouds, would emit UV radiation which is subsequently absorbed and re-emitted by these clouds at IR wavelengths. These same stars would ionise HII regions causing free-free emission and when these stars finally die, undergo a core-collapse supernova (CCSN) accelerating cosmic rays as synchrotron radiation (Condon & Ransom (2016)).

However, this simple star-formation model does not account for the FIRC in starbursts. The power of each cosmic ray electron is proportional to the magnetic energy density and hence proportional to the magnetic field of a galaxy. So a powerful starbursting galaxy (with B 1000μ G) with emit much more powerful cosmic rays that a normal spiral (with B 100μ G) (Condon & Ransom (2016)). The calorimeter model proposed by Völk (1989) to explain this discrepancy. The assumption is that the lifetime of a cosmic ray electron is proportional to the inverse of the magnetic energy density. Thus, cosmic rays from starbursting galaxies can still be powerful, but only radiate a short time, whereas cosmic rays from normal galaxies at a lower power, but for a proportionally longer time. It's from this understanding of radio emission that Bressan et al. (2002) developed their model for modelling radio emission.

The calorimeter model leads to three assumptions; that galaxies are electron calorimeters, UV calorimeters and that synchrotron radiation is the main loss mechanism. These assumptions are unlikely to be true (Lacki et al. (2010)). Consequently, to explain the FIRC, a number of conspiracies have been suggested. For example Lacki et al. (2010) found that in normal galaxies, when electron calorimetry decreases the radio emission, decreasing UV opacity increases the IR emission. For starbursts, Lacki et al. (2010) found that bremsstrahlung, ionization, and IC losses decrease the synchrotron emission, but this is balanced by secondary electrons and the fact that 1.4GHz observation probe lower cosmic ray energies when magnetic field strength increases. In another conspiracy, Bell (2003) found that IR and Radio are not proportional to

the SFR of a galaxy, but they are disproportional to SFR in complimentary ways that cancel each other out.

With the invocation of uncomfortably well-tuned parameters and potential for new physics it is not surprising that the FIRC has enticed a wealth of studies. This has included some models that have ranged from one-zone model of cosmic ray injection (Lacki et al. (2010)), a 1D model of turbulent clumpy star-forming galactic disks (Vollmer et al. (2022)) and a semi-analytic model (SAM) of a single, idealised galaxy (Schober et al. (2022)). In this paper we use Shark, a semi-analytic model of galaxy formation to model the FIRC without invoking any conspiracy. As far as we can assess, there is no other model of galaxy formation that also models the FIRC. This makes Shark the first to directly model a large population of galaxies over cosmic time to include L_{IR} , L_{rad} and consequently qir.

Chapter 2

Methods

2.1 Methods

In this model. we model radio emission as two parts: Free-Free emission and synchrotron emission.

2.1.1 SHARK

SHARK is an open source, semi-analytic model of galaxy formation first presented in Lagos et al. (2018). In this paper we use SHARK v1.1 which is the same presented in Lagos et al. (2018). This version uses the Synthetic UniveRses For Surveys (SURFS) simulation suite (Elahi et al. 2018), specifically the L210N1536 simulation, which uses a Λ CDM (Planck Collaboration et al. 2016) cosmology. That is a Hubble constant of $H_0 = h \times 100 (km/s)/Mpc$, h = 0.6751, matter density of $\Omega_m = 0.3121$, baryon density $\Omega_b = 0.0491$ and dark energy density of $\Omega_{\Lambda} = 0.6879$. The SURFS L210N1536 simulation has a boxsize of $L_{box} = 210 cMpc/h$ and a softening length $\epsilon = 4.5 ckpc/h$. Each box contains $N_p = 1536^3$ dark matter particles each with a mass of $m_p = 2.21 \times 10^8 M_{\odot}/h$. The SURFS simulation suite allows for studies of galaxies with stellar masses above $10^7 M_{\odot}$. This simulation produces 200 snapshots logarithmically arranged from z = 24 - 0. This corresponds to a time between snapshots of $\approx 6 - 80 Myr$.

Halos, subhalos and their properties are identified using VELOCIRAPTOR (Elahi et al. 2019a, Cañas et al. 2019). VELOCIRAPTOR first identifies halos using a 3D friend-of-friend (FOF) algorithm. It also applies a 6D FOF with velocity dispersion to remove spuriously linked objects (like early stage mergers). This 3D FOF structure corresponds to the halo. It then identifies particles that have a local velocity distribution significantly different from the smooth background halo. It runs a phase-space FOF on these particles to identify the subhalos. SURFS only considers halos with $/rm \ge 20$ dark matter particles.

Merger tress are then constructed using TreeFrog (Elahi et al. 2019b). At it's most basic, TreeFrog is a particle correlator that relies on particle IDs being continuous across snapshots. The merger tree is constructed forward in time, indentifying the optimum link between progenitors and descendants. TreeFrog searches up to four snapshots to identify optimal links.

VELOCIRAPTOR and TREEFROG provide the subhalo and merger tree catalogues, respectively, which provide the basis from which galaxies are evolved. SHARK evolves these galaxies across snapshots using a physical model. The physical model used here is fully described by equations 49-64 in Lagos et al. (2018). Before this evolution takes place, the merger trees undergo a post processing treatment.

First 'interpolated' subhalos are inserted between snapshots of the current subhalo and its descendants. This interpolated subhalo has the same properties as the progenitor subhalo. This ensures continuity of galaxy evolution across snapshots. Since TREEFROG searches up to four snapshots for primary links, it is possible that a descendant subhalo is not present in the following snapshot, thus causing discontinuity. The interpolated subhalos solve this problem and enforces continuity of galaxy formation.

Second, the merger tree is checked to make sure that the mass of each halo is strictly equal to larger than the halo mass of its most massive progenitor. This ensures a matter accretion rate onto halos is always ≥ 0 .

Third, the central subhalos are found. At the z=0 snapshot, the most massive subhalo of every existing halo is defined as the central subhalo. The main progenitor of this central subhalo is then defined as the central subhalo of their repspective halos in the next snapshot and so on back through time. Other subhalos that are not the main progenitor or the most massive subhalo of an existing halo are designated satellite subhalos.

Finally, halos that first appear in the catalogue (the ones without a progenitor) are identified. The central subhalo of this halo is assigned a halo gas reservoir of mass $\Omega_{\rm b}/\Omega_{\rm m} \times M_{\rm halo}$ (where $M_{\rm halo}$ is the mass of the halo). Gas cooling is ignited when halo gas mass is > 0, and so a cold gas disk is formed. After each snapshot, the galaxies that are hosted by a subhalo are transfered to their descendants and evolved. If a halo is formed without a progenitor and is a satellite subhalo it is defined as a dark subhalo, is not assigned any gas and does not evolve.

Within Shark there are three different types of galaxies; Centrals, which are the central galaxy of the central subhalo, Sattellites, which are the central galaxy of satellite subhalos and Orphans, which are the central galaxy of a defunct subhalo. A defunct subhalo is one which has merged onto another subhalo and is not the main progenitor. From these definitions a central subhalo can have only one central galaxy, but many orphan galaxies, but a satellite subhalo can only have one satellite galaxies.

In this paper we scale up the total volume of each snapshot by utilising 64 independent subvolumes. Each sub-volume is a box with $V_{\rm box} = 210^3 [{\rm cMpc/h}]^3$ independently evolved from other sub-volumes. This means the total volume under consideration is $V_{\rm total} = 64 \times 210^3 [{\rm cMpc/h}]^3$ in this paper. By increasing the volume, and therefore number of galaxies, being considered we can construct a more robust population of galaxies from which to derive properties. Shark uses the universal Chabrier IMF (Chabrier 2003).

2.1.2 Viperfish and ProSpect

This work builds on the galaxy emission models presented in Lagos et al. (2019). This includes using the Spectral Energy Distribution (SED) created for FIR-FUV emission of Shark galaxies. These SEDs are created using Prospect (Robotham et al. 2020) and Viperfish (Lagos et al. 2019).

PROSPECT is a generative SED package that allows for the creation of SEDs from known star formation and metallicity histories. It can also be used to fit observed SEDs VIPERFISH is an intermediary tool that connects Shark with Prospect. Viperfish extracts star formation histories (SFH) and metallicity formation histories (ZFH) from Shark and translates them to Prospect for SED generation.

PROSPECT was designed to be user friendly with high levels of flexibility in the SFH and ZFH inputs. PROSPECT can make use of the Bruzual & Charlot (2003) or Vazdekis et al. (2016) stellar synthesis libraries and uses a Chabrier (2003) IMF for both libraries. This gives PROSPECT a broad spectral range available for SED generation from stellar synthesis library that is well understood by the astrophysics community. In this paper we use the Bruzual & Charlot (2003) stellar synthesis library and the Chabrier (2003) IMF. This is the same stellar synthesis library used to produce SEDs in Lagos et al. (2019) and the same IMF used in SHARK (Lagos et al. 2018).

PROSPECT's flexibility in its SFH and ZFH inputs makes it ideal for use with SHARK. In PROSPECT, SFH and ZFH can be arbitarily complex; internal interpolation schemes map the provided inputs onto the discrete library of temporal evolution available. Note that if the timesteps within the provided ZFH and SFH are too large this can lead to large uncertainties in the resulting emission, particularly UV. In SHARK, the UV emission is accurately predicted as the time-steps are sufficiently fine (Lagos et al. 2019).

PROSPECT generates SEDs from SHARK by using a simplified fiducial model of dust processing. The light is first attenuated by the Charlot & Fall (2000) dust model. The dust is assumed to be in two phases; birth clouds (BC) and diffuse ISM. The absorption curves for the BCs and ISM are defined in Equations 2.1 and 2.2.

$$\tau_{\rm ISM} = \hat{\tau}_{\rm ISM} (\lambda/5500\text{Å})^{\eta_{\rm ISM}}, \tag{2.1}$$

$$\tau_{\rm BC} = \tau_{\rm ISM} + \hat{\tau}_{\rm BC} (\lambda/5500\text{Å})^{\eta_{\rm BC}}, \tag{2.2}$$

where $\tau_{\rm BC}$ and $\tau_{\rm BC}$ are the optical depths at 5500Å for BC and the diffuse ISM respectively, $\tau_{\rm BC}$ and $\tau_{\rm ISM}$ are the attenuation due to BC and ISM respectively. $\eta_{\rm ISM}$ and $\eta_{\rm BC}$ are the spectral slopes for the two attenuation curves.

Light from different stellar age populations are attenuated by dust differently. Birth clouds contain stellar populations younger than 10 Myr and older stars are not present in birth clouds. Hence, light from younger stars has its light attenuated by both Eqns 2.1 and 2.2 while older

stars are only attenuated by Eqn 2.2.

For the calculation of $\tau_{\rm ISM}$, $\tau_{\rm BC}$ and $\eta_{\rm ISM}$ it is necessary to calculate the surface density of dust $(\Sigma_{\rm dust})$. In Shark the calculation of $\Sigma_{\rm dust}$ is done separately for disks and bulges. For disks the average $\Sigma_{\rm dust}$ is given by:

$$\Sigma_{\text{dust,disk}} = \frac{M_{\text{dust,disk}}}{2\pi r_{50,d} l_{50}} \tag{2.3}$$

where $M_{\text{dust,disk}}$ is the dust mass in the disk, $r_{50,d}$ is the half-gas mass radius of the disk (both of these quantities are outputs in the HDF5 file from Shark) and l_{50} is the projected minor axis. Here $l_{50} = \sin(i) * (r_{50,d} - r_{50,d}/7.3 \text{ where } i \text{ is the inclination.} 7.3 \text{ is the scaleheight to sclelength observed relation in local galaxy disks (Kregel et al. 2002). A galaxy's inclination is determined by the host subhalo angular momentum vector or is randomly chosen for orphan galaxies (Chauhan et al. 2019).$

For bulges the average Σ_{dust} is given by:

$$\Sigma_{\text{dust,bulge}} = \frac{M_{\text{dust,bulge}}}{2\pi r_{50,d}^2}$$
 (2.4)

where $M_{dust,bulge}$ is the dust mass in the bulge and $r_{50,b}$ is the half mass radius of the bulge. Inclination is not important in bulges because we assume they are spherically symmetric. For attenuation due to the diffuse ISM, Lagos et al. (2019) use the work of Trayford et al. (2017) and Trayford et al. (2019).

Trayford et al. (2017) computed the attenuation for each galaxy in the EAGLE hydrodynamical simulation suite using SKIRT (a software package that can be used for simulating radiation transfer in galaxies). These curves were later parameterised to the Charlot & Fall (2000) model (Trayford et al. 2019). From these parameterised curves the median and $1 - \sigma$ relationship between $\tau_{\rm ISM}$, $\eta_{\rm ISM}$ and $\Sigma_{\rm dust}$ were derived. Using these parameterised curves and the $\Sigma_{\rm dust}$ from Shark as calculated in Eqns 2.3 and 2.4 $\tau_{\rm ISM}$ and $\eta_{\rm ISM}$ are calculated. These values are then perturbed by sampling from a gaussian disrtibution with width σ , where σ is the 16th – 84th percentile as predicted by Trayford et al. (2019).

Lagos et al. (2019) follows the Lacey et al. (2016) model for birth cloud attenuation. This model assumes that the birthcloud optical depth scales with the gas metallicity and gas surface density of the cloud, but substitutes the metal surface density with dust surface density of the clouds:

$$\tau_{\rm BC} = \tau_{\rm BC,0} \left(\frac{f_{\rm dust} Z_{\rm gas} \Sigma_{\rm gas,cl}}{f_{\rm dust,MW} Z_{\rm o} \Sigma_{\rm MW,cl}} \right)$$
 (2.5)

where $f_{\rm dust} = M_{\rm dust}/M_{\rm Z}$, (dust-to-metal mass ratio), $\tau_{\rm BC,0} = 1$, $\Sigma_{\rm MW,cl} = 85 \rm M_{\odot} pc^{-2}$, $Z_{\rm O} = 0.0189$ and $f_{\rm dust,MW} = 0.33$ so that a typical sprial galaxy, $\tau_{\rm BC} \approx \tau_{\rm BC,0}$ (matching the result from Charlot & Fall (2000) and Kreckel et al. (2013)).

The cloud surface density is $\Sigma_{\rm gas,cl} = \max[\Sigma_{\rm MW,cl}, \Sigma_{\rm gas}]$ where $\Sigma_{\rm gas}$ is found using Eqs. 2.3 and 2.4. Galaxies in the local group have a giant molecular cloud with constant gas surface

density close to the value of $\Sigma_{MW,cl}$ (Krumholz 2014). This gas surface density is independent of galactic environment and has been found in galaxies ranging from metal-poor dwarfs to molecule-rich spirals. Outside of the local group, the giant molecular cloud surface density increases to maintain equilibrium pressure with increasing ISM pressure. Thus, $\Sigma_{\rm gas,cl} \approx \Sigma_{\rm gas}$ in these environments (Krumholz et al. 2009). A physical limit of $\tan_{\rm BC} \ge \tau_{\rm ISM}$. For birth blouds, Lagos et al. (2019) adopts $\eta_{\rm BC} = -0.7$, being the default value from Charlot & Fall (2000). Once the light has been attenuated by the dust model, it must be re-emitted at longer wavelengths.

Once the light has been attenuated by the dust model, it must be re-emitted at longer wavelengths. To do this, PROSPECT adopts the empirical templates from Dale et al. (2014) (henceforth the Dale model).

The Dale model has been in development since 2001 and consists of observationally constrained templates to model the complete SED of normal star-forming galaxies (SFG). It is an empirical model largely based on the observations of the local Universe and was initially developed for SFGs and only for the infrared SED (between $3-1,100\mu m$ Dale et al. (2001). Based largely on the work of Désert et al. (1990), the Dale model combines the contributions to the SED from large dust grains, very small dust grains and polycyclic aromatic hydrocarbons (PAH) heated by radiation fields of intensity U, with U having a range of $0.3-10^5$. Here U measure multiples of the local interstellar radiation field in the Solar Neighborhood; U = 1 corresponds to the local interstellar radiation field in the Solar Neighborhood.

The Dale model adopts the same large dust grain emission profile as Désert et al. (1990) using greybody emission, the amplitude and wavelength of which varies with U. For very small dust grains the Dale model combines the result of Draine & Anderson (1985) and Tran et al. (2001). Draine & Anderson (1985) found the temperature distribution for a variety of graphite grain sizes. They showed that graphite grains of size $0.02\mu m$ will radiate at a single temperature around 20K, smaller graphite grains will radiate over much larger temperature ranges. However, this was only done for a small range of heating intensities. Tran et al. (2001) showed that small dust grain temperature distribution profiles are equivalent to temperature distribution profiles of larger grains. The Dale model combines these two results and employs the approach from Désert et al. (1990) of integration of flux normalisation of the very small grain emission curves to construct very small dust grain profiles for $U = 0.3 - 10^5$.

For PAH's, the Dale model adopts the PAH spectrum model from Désert et al. (1990), but inserts splices of observed mid-IR spectrum from their own observations made using the Infrared Astronomical Satellite (IRAS) and ISOCAM broadband observations. These are further scaled by integrating over the 12 μ m IRAS filter, to construct the emission curve for PAH's.

The large dust grain, very small dust grain and PAH emission profiles are then superimposed upon each other using relative contribution ratio of 64:4.7:4.3 (large dust grains: very small dust grains: PAHs) as per Désert et al. (1990). This results in what is referred to as a 'local' SEDs: synthetic SEDs for a range of dusty environments.

In order to construct a 'global' SED or that of a galaxy the Dale model assumes a power-law

distribution in a given galaxy of dust mass over heating intensity:

$$dM_d(U) \propto U^{-\alpha} dU, 0.3 \le U \le 10^5,$$
 (2.6)

where $dM_d(U)$ is the dust mass heated by a radiation field intensity U and α represents the relative contributions of different local SEDs. By convolving this semi-empirical model with Infrared Space Observatory (ISO) and IRAS broadband filter bandpasses, Dale et al. (2001) was able to reproduce the SED of galaxies observed using ISO and IRAS in the infrared.

For the generation of SEDs using PROSPECT, we adopt an $\alpha = 3$ for the diffuse ISM and $\alpha = 1$ for birthclouds. These values approximately correspond to effective dust temperatures of 20 - 25 K for the ISM and 50 - 60 K for birth clouds. We also assumed no AGN emission is present (which is left for future work).

By adding the attenuated stellar light with the dust emission together we can create the full generative spectrum. This observed frame is then redshifted using the full spectral resolution available. The spectrum is then passed through a number of filters to span the FUV to FIR for the final outputs. These results do not include nebular emission lines and is only based on broad band emission.

In the 2014 update to the Dale model (Dale et al. 2014), the radio part of the SED was added meaning that the Dale model spans from infrared to the radio spectrum. This means that it is possible to create an SED of a star forming galaxy (SFG) from the infrared to the radio using VIPERFISH and PROSPECT as described above. This extention was implemented using the infrared-radio correlation (qir). By assuming a constant qir, it is trivial to find the corresponding radio luminosity from a given infrared luminosity. However, assuming a constant qir is problematic. Recent observational results have brought in to question the rigidity of this constant qir and later in this paper we show that a constant qir cannot be assumed for all SFGs. Consequently we utilise a different method of finding the radio spectrum that is independent of the infrared light produced by the Dale model. This method, which we discuss in the next sections, was developed in Bressan et al. (2002) and Obi et al. (2017) and henceforth will be referred to as the B02 and O17 respectively.

2.1.3 Free-Free Radiation

When a particle is accelerated by an electric field, free-free radiation is produced (Condon (1992)). Free-free radiation (also known as thermal emission) occurs where an electron has its path deflected by the presence of another charged particle (it is so called free-free emission since the electron is free before and after the interaction with the particle) (Condon & Ransom (2016)).

We model free-free radiation to be proportional to the production rate of Lyman continuum photons. Lyman continuum photons are those with a wavelength smaller than 921Åwhich are responsible for completely ionising hydrogen. Young, massive stars produce large amounts of

Lyman continuum photons which, when incident to the hydrogen gas region surrounding a star completely ionises the hydrogen present. This then leads to an abundance of electrons and positive ions in these regions (known as HII regions or Strömgen Spheres (Strömgren (2013))) hence many electron-ion interactions can take place leading to free-free radiation (Condon & Ransom (2016)). This is the approach used in Bressan et al. (2002) and Obi et al. (2017) which is based on the work of Rubin (1968) and Condon (1992).

The production rate of Lyman continuum photons Q_H is commonly expressed in Eq. 2.7.

$$Q_{\rm H} = \int_0^{\lambda_0} \left(\frac{\lambda L_{\lambda}}{\rm hc}\right) d\lambda, \tag{2.7}$$

where λ_0 is the Lyman limit, 921 Å, L_{λ} is the composite SED of a galaxy in ergs⁻¹ Å⁻¹, h is Planck's constant and c is the speed of light. L_{λ} is sourced from SEDs created using the PROSPECT and VIPERFISH models as discussed in section 2.1.

Rubin (1968) (henceforth Rubin) was the first to quantify the relation between free-free radiation and the number of Lyman continuum photons. While the proof of which will not be reproduced here, Rubin starts by considering that the production rate of Lyman continuum photons will equal the destruction rate of Lyman continuum photons. Lyman continuum photons are destroyed during the photo-ionisation processes of ionising hydrogen. Rubin combines this creation-destruction equilibrium with the expression for free-free radiation total flux density from Oster (1961) and Mezger & Henderson (1967). A key assumption of the free-free radiation total flux density from Oster (1961) and Mezger & Henderson (1967) is that the frequency is large enough to ignore self-absorption.

This leads to the relation that approximates total flux density of free-free emission as a function of the number of Lyman continuum photons, temperature, frequency and distance. There is also a self-absorption term present in this equation, however as Rubin notes, for most galaxies it is possible to ignore self-absorption.

Condon (1992) would later re-express this approximation to be between luminosity, production rate of Lyman continuum photons, temperature and frequency, which is given in Eq. 2.8, In this paper, we model Free-Free radiation in the same way as in B02 and O17 in Eq. 2.8. This model is well understood theoretically and was developed by in Rubin (1968) and Condon (1992).

$$\frac{L_{\rm ff}}{W H z^{-1}} = \frac{Q_{\rm H}}{6.3 \times 10^{32} {\rm s}^{-1}} \left(\frac{T}{10^4 {\rm K}}\right)^{0.45} \left(\frac{\nu}{{\rm GHz}}\right)^{-0.1}$$
(2.8)

Equation 2.8 is used in this model to calculate the free-free radiation and is identical to that used in O17 (See Equation 5 in that paper). Note that this equation is of the same form of the equation used to model free-free radiation in B02, but uses a different constant in the denominator of the production rate of Lyman continuum photons; B02 and we use $6.3 \times 10^{32} s^{-1}$ and B02 ses $5.495 \times 10^{32} s^{-1}$. (See Equation 1 in B02). B02 used their own simulation model of HII regions to calculate an average relation at 1.49 GHz to find $5.495 \times 10^{32} s^{-1}$. In this paper,

we elect to use $6.3 \times 10^{32} \text{s}^{-1}$ since it comes from a purely theoretical understanding of free-free radiation.

We assume a constant temperature of 10⁴K which aligns with observations of HII regions (Anderson et al. 2009).

2.1.4 Synchrotron Emission

When a particle is accelerated by a magnetic field magnetobremsstrahlung (magnetic braking radiation) is produced (Condon & Ransom (2016)). The characteristic of the radiation produced can change depending on the speed of the electron being accelerated. Electrons travelling at ultra-relativistic speeds produce synchrotron radiation (also known as non-thermal radiation), which can account to up to 90 percent of light from SFGs at radio wavelengths (Condon (1992)). The physics behind synchrotron radiation is not as well understood as free-free radiation. As with free-free emission, we take the approach taken by B02 and O17. Relativistic electrons accelerated by core-collapse supernovae (CCSN) into the ISM is used as the dominant mechanism of synchrotron radiation with a minor contribution from supernova remnants (SNR) also included. Synchrotron radiation is calculated through Eq. 2.9, which identical to equation 17 in B02.

$$\frac{L_{sync}(\nu)}{[10^{23}WHz^{-1}]} = \left[E^{SNR} \left(\frac{\nu}{1.49GHz} \right)^{-0.5} + E^{EI} \left(\frac{\nu}{1.49GHz} \right)^{-\alpha} \right] \times \frac{\nu_{CCSN}}{[yr]}, \quad (2.9)$$

where E^{SNR} is the energy contribution from SNR, E^{EI} is the energy of electrons injected per SN event, α , the radio slope from electrons injected per SN event and ν_{CCSNe} is the rate of CCSN. In this equation we assume that E^{SNR} , E^{EI} and α are constants and take the values $E^{SNR} = 1.16 \times 10^{22} \text{WHz}^{-1}$, $E^{EI} = 1.82 \times 10^{23} \text{WHz}^{-1}$ and $\alpha = 0.816$. The values these constants take are derived empirically and differ from those used in the B02 and O17; we change the rate of CCSN within the Galaxy to match the IMF of Shark when deriving these constants. The derivation and assumptions made within remain the same as that in B02. Later in this section, we present the brief derivation of these constants for completeness.

In Equation 2.9 $\nu_{\rm CCSN}$ is not assumed to be constant, but instead calculated from the adopted IMF:

$$\frac{\nu_{\text{CCSNe}}}{[\text{yr}^{-1}]} = \frac{\alpha_{\text{CCSNe}}}{[\text{M}_{\odot}^{-1}]} \times \frac{\text{SFR}}{[\text{M}_{\odot}/\text{yr}]}.$$
 (2.10)

In this paper, we adapt the model to directly calculate ν_{CCSNe} for each galaxy in Shark. ν_{CCSNe} is modelled as directly dependent on the Star-formation rate (SFR) of a galaxy through Equation 2.10 where α_{CCSNe} is the fraction of stars that undergo CCSNe per unit solar mass formed. It is a common assumption that the stars that eventually undergo CCSNe exist within the

mass range of 8 M_{\odot} \lesssim M \lesssim 50 M_{\odot} (Heger et al. (2003), Ando et al. (2003), Nomoto (1984), Tsujimoto et al. (1997)). Above this maximum mass, stars undergo hypernova and causing Gamma Ray Bursts (Van den Heuvel & Yoon 2007). Consequently α_{CCSNe} can be expressed as it is in Equation 2.11 where $\psi(M)$ IMF.

$$\alpha_{CCSNe} = \frac{\int_{8M_{\odot}}^{50M_{\odot}} \psi(M)dM}{\int_{0.1M_{\odot}}^{100M_{\odot}} M\psi(M)dM}$$
(2.11)

As discussed in Section 2.1, we use the use the stellar synthesis library from Bruzual & Charlot (2003) when craeting the SEDs using PROSPECT. The Bruzual & Charlot (2003) stellar synthesis library has a mass range of $0.1 M_{\odot}$ to $100 M_{\odot}$ (hence the mass range chosen in the denominator of Equation 2.11). Both PROSPECT and the Bruzual & Charlot (2003) stellar synthesis library make use of the Chabrier (2003) IMF, where $A = 0.158 (\log M_{\odot})^{-1} \text{ pc}^{-3}$, $m_c = 0.079$, $\sigma = 0.69$ and $B = 4.43 \times 10^{-2} (\log M_{\odot})^{-1} \text{ pc}^{-3}$.

$$\psi(\log(M)) = \begin{cases} A * e^{[-(\log(M) - \log(m_c))^2/2\sigma^2]} & M \le 1M_{\odot} \\ B * M^{-1.3} & M > 1 \end{cases}$$
 (2.12)

where A = 0.158 $(\log M_{\odot})^{-1}$ pc⁻³, $m_c = 0.079$, $\sigma = 0.69$ and B = 4.43×10^{-2} $(\log M_{\odot})^{-1}$ pc⁻³. For a Chabrier (2003) IMF, and using the mass limits above, $\alpha_{\rm CCSNe} = 0.011 {\rm M}_{\odot}^{-1}$.

The derivation of constants in 2.9 is as follows. The B02 derives the total synchrotron emission in our galaxy using the result of Berkhuijsen (1984). Berkhuijsen (1984) found the total synchrotron radiation observationally from our Galaxy at 408 MHz: $L_{0.408 \rm GHz} = 6.1 \times 10^{21} \rm WHz^{-1}$. Assuming a radio slope 0f $\alpha = 0.8$ we convert this to the total synchrotron luminosity at 1.49 GHz: $L_{1.49 \rm GHz} = 2.13 \times 10^{21} \rm WHz^{-1}$. It is possible to then find the average synchrotron luminosity per supernova event. $E^{\rm sync}$:

$$E^{sync} = \frac{L_{1.49GHz}}{\nu_{CCSN,MW}} = 1.24 \times 10^{23} W Hz^{-1}$$
 (2.13)

where $\nu_{\text{CCSN,MW}}$ is the rate of CCSN in the Milky Way Galaxy. $\nu_{\text{CCSN,MW}}$ is assumed to be constant and we take $\nu_{\text{CCSN,MW}} = 0.011 \text{yr}^{-1}$ which is calibrated with the Chabrier IMF used in Shark. B02 and O17 assumes a $\nu_{\text{CCSN}} = 0.015 \text{yr}^{-1}$ which appears to be from Cappellaro & Turatto (2001) and uses a Salpeter IMF. It is this difference in $\nu_{\text{CCSN,MW}}$ that results in the different constants used in this paper than those used in B02 and O17.

An electron's lifetime and the luminosity of a galaxy are both dependent on the magnetic and radiation density fields. Different environments can have differences in these density fields. It is an open question that the FIR/radio correlation is seemingly unaffected by these differences. In order to reconcile this, the Bressan model assumes that the lifetime of synchrotron electrons is much smaller than the fading time of CCSN rate. The Bressan model also assumes that the cosmic ray electrons are injected during the adiabatic phase of SN explosions. On these timescales, the bolometric synchrotron luminosity is

$$L_{bol}^{sync} = \int_{0}^{min(T,\tau_s^{el})} \nu_{CCSN}(T-t) l^{sync}(t) dt, \qquad (2.14)$$

where t is lookback time, $l^{sync}(t)$ is the synchrotron luminosity of the injected electrons after time t, T is the age of the galaxy and τ_s^{el} is the lifetime of electrons against synchrotron losses. For SFGs, $\tau_s^{el} \ll T$ and the SFR is almost constant over the last Gyr, the integral can become

$$L_{bol}^{sync} = \nu_{CCSN} \int_0^{\tau_s^{el}} \frac{dE}{dt} dt = \nu_{CCSN} E^{EI}$$
 (2.15)

where $l^{\text{sync}}(t) = \frac{dE}{dt}$. Hence, the bolometric synchrotron luminosity scales linearly with nu_{CCSN} with a constant E^{EI} .

Equation 2.15 can also be extended for starburst galaxies. To avoid losses from inverse Comptom scattering, $\tau_{\rm s}^{\rm el}$ must be shorter than that of SFGs (Condon, 1992). On a short time scale, we assume that ${\rm nu_{CCSN}}$ is constant.

This relation can also be extended to specific luminosities. This is shown in B02 and will not be reproduced here. The critical assumptions is that $\nu_{\rm CCSN}$ is constant over electron lifetimes and energy loss is due to synchrotron radiation. B02 also shows that specific luminosity has a very weak dependence on magnetic field and argues that bolometric magnitude has no dependence on magnetic fields. This is because, in integrating from specific to bolometric magnitudes, the limits of integration for frequency match with the limits in energy where the power law electron distribution applies.

The Bressan model also considers the contribution of Supernova Remnants (SNR), noting that other sources provide a negligible contribution. Again we will not reproduce this derivation in detail, but will summarise the key assumptions. Using the surface brightness-diameter (Σ -D) relation from observations made by ??, the authors derive the integrated contribution of a population of SNRs which arise from a burst of instantaneous star formation. Assuming that the lifetime of an SNR is associated with its adiabatic phase (Condon (1992)) and integrating over the past SFR, they then derive the average SNR synchrotron luminosity per SN event to be:

$$E^{SNR} \simeq 0.06 E^{sync}. \tag{2.16}$$

Equation 2.16 tells us that the contribution from SNR makes up about 6 percent of the synchrotron emission. The remaining 94 percent comes from electrons injected into the ISM and accelerated by magnetic fields. As previously derived, $E^{\rm sync} = 1.24 \times 10^{23} {\rm WHz}^{-1}$ and so $E^{\rm EI} = 1.25 \times 10^{23} {\rm WHz}^{-1}$, $E^{\rm SNR} = 0.0795 \times 10^{23} {\rm WHz}^{-1}$.

SNRs have a spectrum which is modelled by $L_{\nu} = \nu^{\alpha_{\rm SNR}}$ where $\alpha_{SNR} = 0.2 - 0.5$. The Bressan model assumes that the radio slope of SNRs is constant at $\alpha_{\rm SNR} = 0.5$, which is less than the characteristic observed slope of the total non thermal emission of normal galaxies ($\alpha_{\rm sync} = 0.8$). In order to compensate for this, the Bressan model assumes that the spectrum for electrons

2. METHODS

injected into the ISM has a radio slope of $\alpha_{\rm EI} \simeq 0.9$ for an overall synchrotron radio slope of $\alpha_{\rm sync} = 0.8$.

We differ in our approach to radio slopes. Like the Bressan model we too assume that $\alpha_{\rm SNR} = 0.5$ but use a $\alpha_{\rm EI} = 0.816$ since this more accurately produces an overall slope of $\alpha_{\rm sync} = 0.8$. Solving Equation ?? gives the result of $\alpha_{\rm EI} = 0.816$.

Implementing the effects of synchrotron self-absorption was considered, however these effects have been shown to be negligible at the brightness temperatures of SFGs (Condon (1992)). Because of this, we do not include the effects of synchrotron self-absorption in this model.

Chapter 3

Results

3.0.1 Local Universe (z = 0)

Comparisons with the GAMA Survey

Figure 3.1 is a comparison plots between the model from Shark and observational data of individual galaxies. The radio data presented here is sourced from the SFG population from Davies et al. (2017) at 1.4GHz. This data is sourced by combining data from the Galaxy And Mass Assembly (GAMA) survey and the Faint Images of the Radio Sky at Twenty cm (FIRST) survey. The IR data is sourced from the cross-matched catalogue from Bellstedt et al. (2020) which uses data from the fourth data release of the Kilo-Degree Survey (KiDS) using the Very Large Telescope (VLT). For simplicity I shall refer to this data as being from the GAMA survey. This data is mostly from the local Universe with redshifts of $0 \le z \le 0.5$.

Davies et al. (2017) uses a robust methodology of removing AGNs from the sample. This methodology is discussed in depth in that paper, but briefly it involves: (i) classifying AGNs using a BPT diagram (Baldwin et al. 1981), (ii) using a colour selection using Wide-field Infrared Survey Explorer (WISE) to remove dust obscured objects and passive galaxies, (iii) removing radio luminous objects with a rest-frame radio luminosity $\geq 10^{23.5} \mathrm{WHz}^{-1}$, (iv) using the Near Infrared (NIR) to Mid-Infrared (MIR) ratio (NIR/MIR) (Seymour et al. 2008), (v) those with an r-band effective radius of > 8 arcsec (Hopkins et al. 2003) and (vi) visually inspecting the spectrum for broad H α emission and two component radio emission.

This rigorous AGN cleaning process means the final sample is scoured of most AGN contamination. The process described above is helped by the low redshift of these galaxies. The galaxies that have survived the AGN cleaning process are then matched with those from Bellstedt et al. (2020) to find their total IR observations.

This data is then compared with Shark. Only galaxies defined as SFG (as determined in the process outlined in section XYZ) at z=0 are compared here. This selection closely matches GAMA's population of galaxies without introduces selection biases. In Figure 3.1, individual GAMA galaxies are shown as dots and the solid blue line represents the median of the Shark galaxies selected here. The blue area around this line in the left-hand plot shows the $1-\sigma$

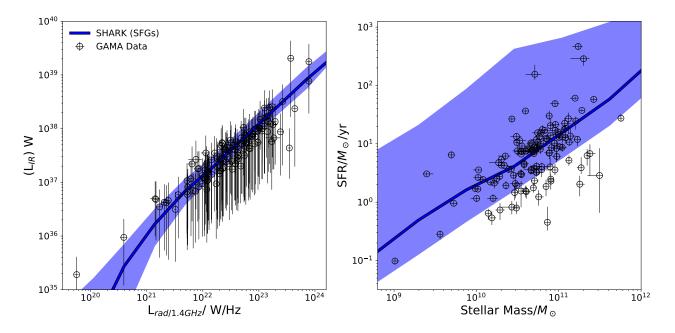


Figure 3.1: Comparison plots with observational data of individual SFGs from the GAMA survey. In both plots, the circles are observational data from the GAMA survey and the blue area is representations from Shark using our radio model. The left plot shows the L_IR vs. $L_rad(1.4GHz)$ for these two populations. In this plot, the area in blue around the median line represents 1- σ uncertainty. The right plot shows SFR vs. total stellar mass where the shaded blue region shows the full extent of the galaxy population, hence the extent of our definition of SFGs. Both plots contain the same galaxies from the GAMA survey and the same subset of galaxies from Shark. In this subset we have chosen all SFGs at z=0, withing the stellar mass range of $8 \le \log(M/M_{\odot}) \le 12$ and with a total infrared luminosity greater than $10^{35}W$.

values of the Shark population. In the right plot, the blue area shows the full extent of the population.

One small caveat is that the GAMA data shows individual galaxies, while the Shark line represents a population of galaxies. It is clear from the Figure 3.1 that there are some galaxies from GAMA that are not within the $1-\sigma$ value of Shark. This does not mean that these galaxies have no counterpart in Shark, just that they appear outside the standard population of Shark galaxies.

The left panel of Figure 3.1 shows L_{IR} to $L_{Rad,1.4GHz}$. This plot is used to show the FIRC, which is present in both the GAMA and SHARK populations. Both populations agree well with each other with all GAMA galaxies being at least within the margin of error from the median SHARK galaxy at the same L_{rad} .

That the Shark re-creates the FIRC is not surprising; the model of radio emission depends on it (Bressan et al. 2002). That Shark re-creates the FIRC in line with independent observations from GAMA is the point here. This is a first confirmation that the radio model used in this paper is reasonable and re-creates observational results.

While the left panel of Figure 3.1 shows a mostly straight line between L_{IR} and $L_{Rad,1.4GHz}$, the median line dips for low luminosity galaxies $L_{Rad,1.4GHz}$ 10^{21} . It is at this point that FIRC appears to break down. This is discussed further in section XYZ of this paper.

The right panel of Figure 3.1 shows the SFR-Stellar Mass plane. This plot is a diagnostic plot showing the veracity of the definition of a SFG used in this paper. This shows that the definition employed here is reasonable, though perhaps a greater definition could have been used. There are some SFGs that appear in GAMA that don't have a counterpart in SHARK and that's due to the restrictive definition employed here. However, overall, it does show that this is a reasonable definition of SFGs.

Radio LFs at z = 0

Figure 3.2 shows the radio luminosity function (LF) at z = 0 at a frequency of 1.4GHz and 150MHz. The radio LF shows the distribution of galaxies across different bins of luminosity. To find the radio LF I first took a histogram of the log_{10} radio luminosities of all galaxies at z = 0. This histogram is taken over a set of bins with equal bin width. This histogram is then divided by the total volume of the simulation (for all 64 snapshots this is $210 \text{Mpc}^3 \times 64$ snapshots) and the bin width of the histogram. The $1 - \sigma$ confidence intervals is then estimated using bootstrapping.

In observations, there can be many sources of error that can affect the results of the radio LF. These can include radio K correction, error in completeness, contamination by AGNs (Novak et al. 2017), cosmic variance (Driver et al. (2022)) and low sample sizes (Bonato et al. (2021b)). In Shark, no K correction is required (Though we do assume a constant spectral index of $\alpha = -0.8$) and the resolution is clearly complete. No AGNs are modelled in the radio luminosities so contamination cannot occur. Any affect from cosmic variance would be very low

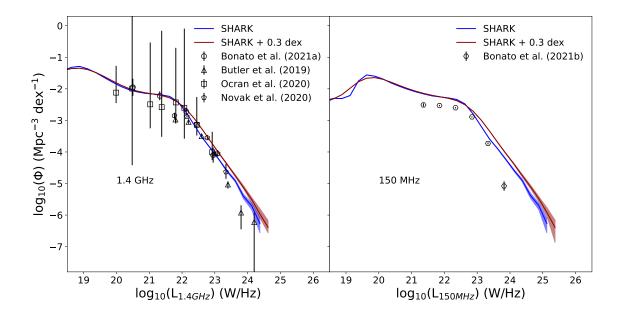


Figure 3.2: The radio LF for all galaxies at z=0 for frequencies 1.4GHz (left) and 150MHz (right). Left: The blue line shows the LF with no convolution while the dark red line shows the LF convolved with a Gaussian with a mean of 0 and a standard deviation of 0.3. Errors have been approximated using bootstrapping and the $1-\sigma$ lines are shown in the shaded regions about the median line. Comparisons are made with observational data of SFGs from Bonato et al. (2021a), Butler et al. (2019) (This data also includes high-excitation sources for which star formation is the likely cause of their radio emission), Novak et al. (2017) and Ocran et al. (2019). Right: Same as the left plot except at 150MHz. Comparisons are made with observational data of SFGs from Bonato et al. (2021b).

since the volume of the sample is $13,440Mpc^3$. Finally, our sample sizes are very large due to the computational ease by which galaxies are modelled.

In order to estimate the effects that the results of these errors can have on the radio LF I convolve the radio LF. During convolution, each individual galaxy luminosity is mulitplied by a random constant, before binning. This constant is chosen from a random Gaussian with a mean of 0 and a standard deviation of 0.3. Convolving luminosities in this way provides an analogue estimation of errors made in observations. This process is the same for both frequencies. The convolved radio LF is shown in red and the non-convolved radio LF is shown in blue.

The result of this is compared with the observational results from Bonato et al. (2021a), Butler et al. (2019), Novak et al. (2017) and Ocran et al. (2019) (for 1.4GHz) and with Bonato et al. (2021b) (for 150MHz). These all show for SFGs, though different methods have been used to determine the presence of AGNs and to quantify the errorbars.

Bonato et al. (2021a) (from which these observational results are sourced) used data from the Westerbork Synthesis Radio Telescope (WSRT) to derive the radio LF. This survey covered an area of $\sim 1.4~\rm deg^2$ and to a redshift of z ~ 3.2 . Radio Loud AGNs are removed using a radio threshold and a radio/SFR threshold. Radio Quiet AGNs are classified using the criterion from Messias et al. (2012). Error bars only show the Poisson error of the sample.

Butler et al. (2019) used the Australia Telescope Compact Array (ATCA) to obserse an area of ~ 25 deg² to a redshift of z ~ 1.3. AGNs are removed using a decision tree in Butler et al. (2018). This decision tree uses: (i) X-Ray Luminosities and Hardness ratios, (ii) SED fitting, (iii) Mid-Infrared colours, (iv) Optical Spectra, (v) Optical colours and (vi) Radio Luminosities, spectral indices and morphologies to decide whether a source contains an AGN or not. Despite this comprehensive method of AGN cleaning, Butler et al. (2018) ackowledges that AGN contamination is still possible within their data. The error bars here represent the standard deviation of the inverse completeness fraction.

Novak et al. (2017) used the Very Large Array (VLA) covering an area of 2 \deg^2 to a redshift of $z \sim 5$. The authors did not remove all AGNs from their sample, rather only radio loud AGNs. The radio loud AGNs were removed with SED fitting. The error bars here are the Poisson error of the sample and does not include errors from radio K correction, sample completeness or radio quiet AGN contamination.

Ocran et al. (2019) used the Giant Meterwave Radio Telescope (GMRT) to measure an area of 1.2 deg^2 , to a redshift $z \sim 5$ at 610MHz. They use the sample from Ocran et al. (2020) and convert to 1.4GHz by assuming a constant spectral index of $\alpha = -0.8$. AGNs are removed using the following diagnostics: (i) a radio cutoff, (ii) the mid-infrared radio ratio, (iii) X-Ray luminosity, (iv) colour-colour diagram and (v) the use of BOSS CLASS and SUB CLASS (Bolton et al. (2012)). Error bars are calculated using confidence tables from Gehrels (1986).

For the 150MHz radio LF, Bonato et al. (2021b) used data from the Low Frequency Array (LOFAR). They use SED fitting to determine AGNs. Error bars are the quadratic sums of the Poisson uncertainty and sample variance.

At z = 0, we find excellent agreement in the radio LF at 1.4GHz. At this frequency, the

non-convolved line appears to agree more with the observational data than the convolved line. Notably, this LF was produced without the need to invoke a top-heavy IMF.

Numerous studies into modelling the UV-FIR have required using a top-heavy IMF (that is, an IMF that produces an overabundance of high mass stars) to reproduce observational results. (Baugh et al. 2005; Lacey et al. 2008; Camps et al. 2016; Cowley et al. 2019; Trčka et al. 2020). The exception to this is Lagos et al. (2019), on which this study builds. Lagos et al. (2019) was able to reproduce the UV-FIR LFs using Shark without invoking a top-heavy IMF. The reasons for this discrepancy between top-heavy and non-top heavy IMF models is discussed in depth in Lagos et al. (2019), but differences in the physical models is the cause of this discrepancy. However as the processes within the physical models are complex and intertwined, a single culprit cannot be found.

This discussion comes with the caveat that as best I can tell, there are no other radio LFs that have been created using SAM except for the Shark model shown here. It's conceivable that innate differences between the modelling of the UV-FIR LF and the radio LF mean that the UV-FIR is more susceptible to changes in the IMF than the radio. But such discussions stray away from the purpose of this thesis. The overall point is this; the model presented here can successfully reproduce the radio LF at 1.4GHz without invoking a top-heavy IMF.

The right panel of Figure 3.2 shows the radio LF at z=0 for 150MHz. Contrary to the left panel, this does not show good agreement with observational results. The non-convolved model agrees with the points form 22.8-23.3, but over predicts the LF at other points. The convolved line does not predict any of the points at all.

Since the Shark model is over predicting compared with observations it is possible to rule out two common causes of disagreement: a top-heavy IMF and AGN contamination in the observational data. The reason a top-heavy IMF is invoked in other studies is because they do no have enough high mass stars to reproduce the light. This causes their LF to under predict compared with observations (the opposite of what is seen here). If the discrepancy in the 150MHz LF is caused by the IMF, it would required a top-light IMF (an IMF that produces a scarcity of high mass stars). Further, the IMF is unlikely to be the cause of this discrepancy since the LF is able to be reproduced at 1.4GHz and in UV-FIR in Shark (Lagos et al., 2019). Further, it is unlikely to be AGN contamination of the data from Bonato et al. (2021b) since Shark is under predicting compared with the data. AGN contamination causes an increase in the luminosity at radio wavelengths. If AGN contamination were a factor in this data, then Shark would be underpredicting the data. Also, cleaning data for AGNs is far easier at low redshifts than at high redshifts, meaning that it's more likely that more AGNs have been removed from the low redshift data here than at higher redshifts.

There are two likely causes of this discrepancy: underestimation of error in the Bonato et al. (2021b) data and spectral index used in Shark. The data from Bonato et al. (2021b) does not include errors due to cosmic variance and low sample sizes, instead using Poisson error (notably, the final point of this data only contains two galaxies). This error is then systemically underestimated in this data set. By including this error it's likely that Shark would better

agree with the data shown here.

The data from Bonato et al. (2021b) was measured at 150MHz so there is no conversion using spectral index. However the model used in Shark is calibrated to 1.4GHz and then converts to other frequencies assuming that SFGs follow a power law distribution with a spectral index of $\alpha = -0.8$. While this assumption is widely spread widely in the literature, there are some suggestions that this might not be the case (Delhaize et al., 2017). Since the observational data appears below the Shark line, this suggests that the spectral index used in Shark is not steep enough. If the spectral index used here is the reason for this discrepancy, then it should be reduced.

It is also worth mentioning that as best that I can tell, Bonato et al. (2021b) is the only paper where the RL is found observationally at 150MHz without assuming a spectral index. Other observations are usually done at 1.4GHz and converted to 150MHz by assuming a spectral index. That Shark disagrees with one source of observational data is neither surprising nor a reason to disregard the Shark model entirely.

Previous studies have also found discrepancies between observational and modelled LFs due to underestimation of the SFR (Somerville et al., 2012). This is discussed at length in section XYZ where it is shown not to be the case.

Overall the Shark RLs agree at 1.4GHz with observations, but do not quite agree at 150MHz. This may be due to many factors including the spectral index used in the Shark model and the underestimation of error in the observational data.

3.0.2 The Distant Universe (z > 0)

Comparisons with Starburting Galaxies

Figure 3.3 shows a comparison plot between Shark and the observational results from Lo Faro et al. (2015). A comparison with Sargent et al. (2010b) is also made. This figure shows is a test of the Shark model in the extreme regime of (U)LIRGs. Error bars for all properties from Lo Faro et al. (2015) is the median error of each of the populations; no individual errors were provided for any property.

Lo Faro et al. (2015) uses a sample first published in Fadda et al. (2010) and is an extension to the analysis done in Lo Faro et al. (2013) using the Spitzer Space Telescope to find the MIR spectroscopy and Hershel Space Observatory for FIR photometry and the VLA at 1.4GHz for radio data. The sample from Fadda et al. (2010) was designed to characterise the properties of sources which contribute to the Cosmic Infrared Background (CIB). It has been shown that the main contributors to the CIB are sub mJy sources at 24 μ m with a main peak at z 1 and a secondary peak at z 2 (Lagache et al. 2005). At these redshifts, sub-mJy IR sources would be LIRGs at z 1 and ULIRGs at z 2, which is the definition that Fadda et al. (2010) uses in classifying the galaxies.

In the observational data used here, LIRGs are classified as having $S(24\mu m)$ 0.2 – 0.5 mJy at z = 0.76 - 1.05. ULIRGs are classified as having $S(24\mu m)$ 0.14 – 0.55 mJy at z = 1.75 - 2.4.

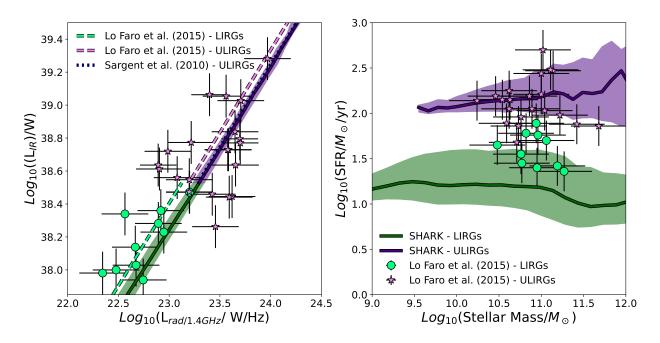


Figure 3.3: Comparison plots with observational data of individual (U)LIRGs from Lo Faro et al. (2015). The circles are the observational data from Lo Faro et al. (2015) of LIRGs and stars are ULIRGs. The blue area is LIRGs and the red area is ULIRGs from the SHARK with radio luminosity calculated using our model. The right plot shows L_IR vs. $L_rad(1.4GHz)$ and the left plot shows SFR vs. total stellar mass. LIRGs from SHARK have been selected as having a total $10^{11}L_{\odot} < L_{IR} < 10^{12}L_{\odot}$ and ULIRGs have been selected as having a total $10^{12}L_{\odot} > L_{IR}$. These are the same parameters that constrain the (U)LIRGs found in Lo Faro et al. (2015). No selection with redshift have been made of the SHARK galaxies.

This differs from the usual definition of LIRGs having IR luminosity $10^{11} - 10^{12} L_{\odot}$ and ULIRGs having IR luminosity of $10^{12} - 10^{13} L_{\odot}$, which is the definition I use in Shark. It also rules out any Hyper Luminous InfraRed Galaxies (HyLIRGs) and Extremely Luminous Infrared Galaxies (ELIRGs) which are usually defined as having an IR luminosity of $10^{13} - 10^{14} L_{\odot}$ (Cutri et al. 1994) and $> 10^{14} L_{\odot}$ (Tsai et al. 2015) respectively, however no galaxies in this sample has an IR luminosity $> 10^{13} L_{\odot}$.

The stellar mass of the galaxies are determined using SED fits using two different codes: Fadda et al. (2010) and Lo Faro et al. (2013) use HYPERZ (Bolzonella et al. (2000) while Lo Faro et al. (2013) and Lo Faro et al. (2015) uses GRASIL (Silva et al. 1998). (Further, more specific details of how these fits are completed are present in the three separate papers). The stellar mass shown here is that which has been found using HYPERZ. The reason for this is because in modelling the radio frequencies of the SEDs, GRASIL invokes the Bressan model; the same model that is used here to model radio emission (Vega et al. (2005)). While Lo Faro et al. (2013) does use GRASIL to find the stellar mass using FIR photometry and MIR spectroscopy and hence would not invoke the Bressan model in using GRASIL, to avoid any potential circularity, I use the results from HYPERZ. Lo Faro et al. (2013) finds that the GRASIL finds higher stellar mass than HYPERZ, Figure 3.3 shows that the stellar mass ranges are in broad agreement with that found in SHARK (The differences in the GRASIL and HYPERZ stellar masses are not that significant on these scales in any case).

GRASIL is also used derive the SFR in Lo Faro et al. (2013) and Lo Faro et al. (2015), however the SFR is also calculated using the SFR- $L_{\rm IR}$ relation from Kennicutt Jr (1998). This relation is shown in Equation 3.1.

$$SFR = \frac{L_{FIR}}{5.8 \times 10^9 L_{\odot}} \tag{3.1}$$

Equation 3.1 was calculated for a Salpeter IMF (Salpeter 1955), however this was rescaled to a Chabrier IMF in Lo Faro et al. (2013). For the same reasons discussed above, I adopt the SFR found using Equation 3.1.

One of the results of Lo Faro et al. (2013) was to show that, at least for (U)LIRGs, that the Kennicutt Jr (1998) produces a SFR 1.7 - 2.5 higher than the result from GRASIL. This is because the Kennicutt Jr (1998) method include significant "cirrus" emission from intermediate-age stellar populations which contribute to dust heating. Figure 3.3 shows that Shark produces SFR that are lower than that calculated using the relation from Kennicutt Jr (1998) for LIRGs, but, on average, produce SFRs that are comparable for ULIRGs.

Lo Faro et al. (2015) also use GRASIL to fit for L_{IR} and $L_{1.4GHz}$, but present that which is calculated directly from observations. I use here the results direct from observations, without invoking GRASIL.

As an aside, Lo Faro et al. (2015) finds excellent agreement with the $L_{1.4GHz}$ as derived from GRASIL and the observations. From this it is not surprising that SHARK finds broad agreement with these observations using the same model.

AGNs are identified in Fadda et al. (2010) and excluded in Lo Faro et al. (2013) using several different AGN indicators including: (i) broad and high ionisation lines for those galaxies with optical spectra, (ii) SED fitting, (iii) X-Ray brightness (for sources with X-Ray observations), (iv) strength of 6.2 μ m Polycyclic Aromatic Hydrocarbons (PAH) features and (v) optical morphology. The overall sample is crudely luminosity selected and results in 31 (U)LIRGs in the sample.

Sargent et al. (2010b) studied ULIRGs out to z < 2 using the VLA-COSMOS "Joint" Catalogue. This is an IR selected sample and they found that qir = $2.672^{+0.121}_{-0.121}$ that is independent of redshift. This includes 1692 star-forming ULIRGs and 3004 'IR bright' sources. AGNs are removed from this sample using a method derived in Smolčić et al. (2008). Briefly, this method uses the rest frame (u – K) colours and relies on the correlation between these colours and the BPT diagram.

The Shark population of galaxies are chosen to following the definitions of (U)LIRGs given in Fadda et al. (2010): LIRGs having L_{IR} between $10^{11} - 10^{12}L_{\odot}$ and ULIRGs have $L_{IR} > 10^{12}L_{\odot}$. Note that unlike Lo Faro et al. (2015), no redshift selection has been taken for (U)LIRGs. Including a redshift selection did not drastically change the results shown in Figure 3.3 and so a broader population of galaxies was chosen in order to compare with here.

The left panel of Figure 3.3 shows the L_{IR} to $L_{1.4GHz}$ relation and hence the existence of the FIRC in (U)LIRGs. It shows that the medians from Lo Faro et al. (2015) does not necessarily agree well with Shark, but the results from Sargent et al. (2010b) agrees very well with Shark (So much that it is hard to see!)

The reason for this discrepancy between Lo Faro et al. (2015) and Sargent et al. (2010b) is sample size. Lo Faro et al. (2015) only has a total sample size of 31 galaxies whereas Sargent et al. (2010b) has a sample size of over 3000. This is a dramatically more robust and statistically signicant sample. Hence, it is not concerning that the median of Lo Faro et al. (2015) does not agree with Shark and heartening that the median of Sargent et al. (2010b) does. That being said, Sargent et al. (2010b) does not offer any other characteristic measurements, like stellar mass or SFR, whereas Lo Faro et al. (2015) does.

It is important to remember that the results from Shark show the median and uncertainty of an entire population, whereas the points from Lo Faro et al. (2015) shows the results of individual galaxies. Thus, that there are individual points away from the Shark median is to be expected. In any case, nearly all of the points are within their margin of error of the median from Shark.

Another reason why the median of Lo Faro et al. (2015) is higher than that of Shark is selection biases within the observational sample. The Fadda et al. (2010) only selected galaxies with available optical and near IR data with which properties could be derived. As these galaxies are at high redshift, they would require a high optical luminosity. Such galaxies might be biased towards higher IR and radio luminosities, even within the (U)LIRG regime. This would lead to a qir that is the same slope, but shifted across towards higher L_{IR} .

The main takeaway from the left figure is that the FIRC exists in (U)LIRGs within SHARK, in

line with the observations from Sargent et al. (2010b) and Lo Faro et al. (2015).

The right panel of Figure 3.3 shows the SFR-stellar mass relation. This shows that for SHARK, (U)LIRGs do not evolve significantly over the SFR-stellar mass relation (though no evolution was expected). It also shows that (U)LIRGs in SHARK produce a high SFR despite no SFR selection was made in the population.

This shows excellent agreement between Shark and Lo Faro et al. (2015) in terms of ULIRGs, but less agreement in terms of LIRGs. This disagreement is not fatal; I am comparing individual galaxies to a population after all so some scatter is expected. Further, most of the LIRGs are within the $1-\sigma$ range of the Shark population, but it is worth discussing all the same This could be due to Kennicutt Jr (1998) systemically producing higher SFR, as Lo Faro et al. (2013) found. This would increase the SFR of each of the LIRGs beyond normal leading to the disagreement that is seen here. That ULIRGs do agree with Shark is also because of this systemically high SFR and that the definitions used to classify (U)LIRGs used here are different. There are some ULIRGs with IR luminosity < $10^{11}L_{\odot}$ in the observational sample. These galaxies are excluded by definition in the Shark ULIRGs. Hence, compared with the Shark ULIRGs, the observational ULIRGs will have a lower IR luminosity. In estimating the SFR as proportional to IR luminosity, this lower IR luminosity would produce a lower SFR. This affect appears to cancel out the systemic over estimation of SFR that the Kennicutt Jr (1998) method has.

It is also indicative of a sample that is biased towards high IR galaxies. If the sample does contain uncharacteristically high IR, then this would produce a high SFR by definition when using a $L_{IR} - SFR$ relation.

Further, Fadda et al. (2010) showed that the sample was biased towards galaxies that contain PAH features. PAHs trace cold moelcular gas where star formation is optimal and hence are understood to trace star formation (Li 2020). This means this is a sample of galaxies that have a high IR emission and PAH features, meaning they are doubley biased towards galaxies with high SFRs. This bias could cause the increase in LIRGs seen here.

While there are differences between the observations from Lo Faro et al. (2015) and Shark they are really academic. The median Shark result is within the margin of error for most of the observations. The overall point of this figure is that even in the extreme (U)LIRG regime, the Shark model reproduces what is seen observationally. I can therefore have confidence that this model, on average, produces physically meaningful results.

Radio Luminosity Functions at z ¿ 0

Like the left panel of Figure 3.2, Figure 3.4 shows the radio LF at 1.4GHz compared with observations from Bonato et al. (2021a), Butler et al. (2019), Novak et al. (2017) and Ocran et al. (2019) except over multiple redshifts. The same method as outlined in section ?? was followed here. As in Figure 3.2, the blue line shows the radio LF with no convolution and the red line shows the radio LF convolved with a Gaussian with mean = 0 and a standard deviation = 0.3.

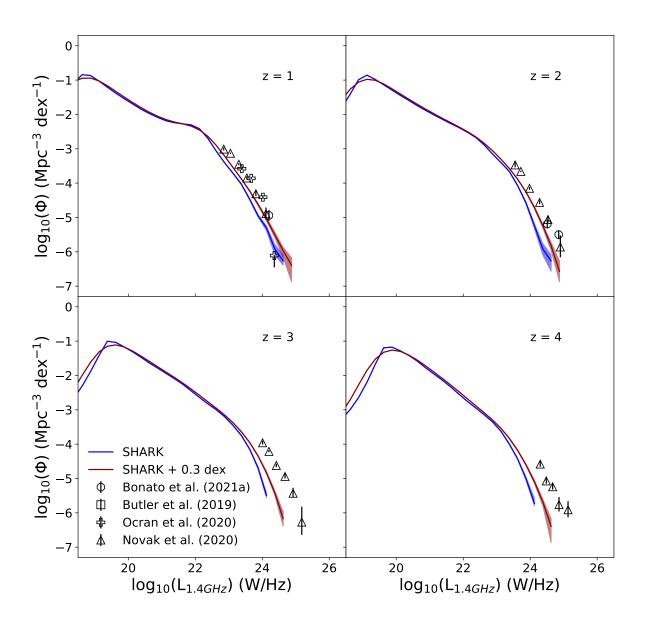


Figure 3.4: The radio LF for all galaxies for different redshifts for frequency of 1.4GHz. Top-left panel shows z=1, top right panel shows z=2, bottom left panel shows z=3 and bottom right panel shows z=4. Much of the details here are the same as in Fig. 3.2; the blue line shows the median with no convolution and the orange shows the same but convolved with a Gaussian with mean =0 and 0.3 standard deviation. Comparisons are made with observational data of SFGs from Bonato et al. (2021a), Butler et al. (2019), Novak et al. (2017) and Ocran et al. (2019). The shaded region shows the $1-\sigma$ area estimated using bootstrapping.

The shaded region shows the $1 - \sigma$ error estimated using bootstrapping.

Figure 3.4 shows that there is less agreement between the Shark model radio LF with observations at higher redshifts. At z=1, the observational data agrees more with the convolved line than the non-convolved line. This is true at z=2 as well. The purpose of this convolved line is to estimate the errors present in these observations like cosmic variance, low sample sizes and AGN contamination which have not been accounted for in the observational data. That the observational data agrees more with the convolved line shows that the regular Shark model is under predicting the radio LF compared with observations.

At z = 3 and z = 4, neither the convolved line nor non-convolved line agrees with the observational data from Novak et al. (2017). At these redshifts, the SHARK model under predicts the radio LF compared with observations, but to a much greater degree than at z = 1 and z = 2.

There are similar trends present in Figure 3.5. Here the radio LF is shown for 150MHz for redshifts 1, 2, 3 and 4. This is compared with observational results from Bonato et al. (2021b). Note that the results from Bonato et al. (2021b) has been binned into larger redshift bins (ie 0.8 < z < 1.0 and 1.0 < z < 1.2) so where applicable, I show two sets of bins that cover the redshift under consideration.

As in Figure 3.4, at z = 1 and z = 2, the observational data agrees more with the convolved Shark model than the non-convolved model. Whereas at z = 3 and z = 4, neither the convolved model nor non-convolved model agrees with observations.

Here I will discuss three possibilities for this discrepancy: that Shark requires a top-heavy IMF, that Shark underpredicts the SFR and that the observations include AGN contamination.

As discussed in section 3.0.1, previous studies into modelling the UV and IR LF have had similar problems in that the models were under predicting the LF compared with observational results. Invoking a top heavy IMF within such models appears to rectify the issue in these studies (Baugh et al. 2005; Lacey et al. 2008; Camps et al. 2016; Cowley et al. 2019; Trčka et al. 2020). I argue that this is not the case in Shark. Shark uses the Chabrier (2003) IMF, which is not top heavy, yet Figure 3.2 shows excellent agreement with observations at = 0. Further Lagos et al. (2019) showed that the UV - IR LFs can be reproduced in Shark using the same Chabrier (2003) IMF. It would require an uncomfortable level of fine tuning to require a top-heavy IMF for z > 0 but not z = 0.

Another possibility which is discussed in the literature is that the SFR in the model is too low (Somerville et al., 2012). This issue is particularly pertinent in Shark as the radio emission is modelled proportional to the SFR (See the Methods section). This possibility is tested in Figure 3.6.

Figure 3.6 shows the SFR function as derived in SHARK. The method of determining the SFR function in SHARK follows closely the process used in determining in the radio LFs. At each redshift, a histogram with equal bin sizes is created from the SFR of all galaxies present. At this point, the histogram undergoes bootstrapping to determine the $1 - \sigma$ uncertainty. The histogram and errors are then divided by the total volume of the simulation (for all 64 snapshots this is $210 \text{Mpc}^3 \times 64$ snapshots) and the bin width.

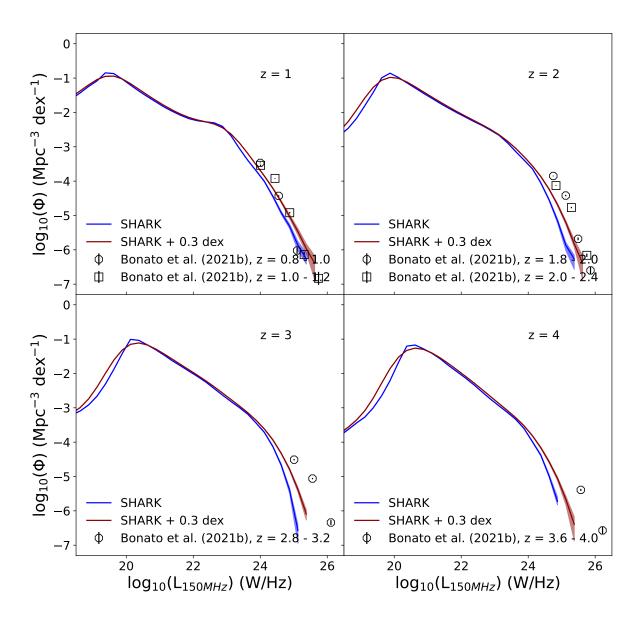


Figure 3.5: The radio LF for all galaxies for different redshifts for frequency of 150MHz. As in Fig. ??, top-left panel shows z = 1, top right panel shows z = 2, bottom left panel shows z = 3 and bottom right panel shows z = 4. Much of the details here are the same as in the right hand panel of Fig. 3.2; the blue line shows the median with no convolution and the orange shows the same but convolved with a Gaussian with mean z = 0 and 0.3 standard deviation. Comparisons are made with observational data of SFGs from Bonato et al. (2021b) The shaded reagion shows the z = 1 area estimated using bootstrapping.

This is compared with observational results from Katsianis et al. (2017). This paper focuses on the evolution of the SFR function in the EAGLE simulation, but to do so derives the SFR function from observations. Marchetti et al. (2016), Mauch & Sadler (2007), Patel et al. (2013), Robotham et al. (2011), Alavi et al. (2016) and Bouwens et al. (2015) (among others used in Katsianis et al. (2017)) derive LFs for their respective tracers (IR, UV, radio etc.). These LFs are then converted to SFRs using relations derived for those tracers. More details of the methods used is present in Katsianis et al. (2017).

What Figure 3.6 shows is that Shark agrees with the SFR derived from observations across different redshifts and different tracer methods. This agreement is in contrast to previous galaxy emission models which under predicted the SFR (Somerville et al., 2012). This therefore refutes the idea that the tensions between Shark and observed radio LFs is due to an underprediction of the SFR.

The third possibility I will discuss for the discrepancy between the Shark and observed radio LFs is that observations contain AGN contamination of what should only contain SFGs. The removal of AGNs is treated differently in each paper (See section 3.0.1 for details). For example, Novak et al. (2017) determines AGN dominated galaxies using a three component SED fit. They also remove radio excess sources by determining the IR derived SFR and construct histograms of $\log_{10}(SFR/radio)$. By the FIRC, the SFR derived from IR and the radio luminosity should be well correlated. Thus they define radio excess sources as being 3σ above the median of these histograms.

Such a method is fine for identifying radio loud AGNs, but as Novak et al. (2017) notes: "We are...not interested in removing all AGN host galaxies from our sample, but only those that show clear evidence of radio emission dominated by an AGN." Hence, AGN contamination is present in these radio LFs. AGN contamination would arbitrarily increase the radio luminosity of each point, which would explain Shark's apparent under prediction in Figure 3.4.

But if AGN contamination is the reason for the tension in Figures 3.4 and 3.5, then why is this not present at z = 0? This comes down to the ease for which AGNs can be identified at low redshifts compared with high redshifts. For example the BPT diagram used to identify AGNs in the GAMA data is unusable at z > 0.4. This is because emission lines required to construct a BPT diagram are shifted into the IR spectrum where deriving spectra becomes increasingly difficult (Trouille et al. (2011)). To that end there is a difficulty in attaining a clean sample of SFGs at high redshifts that is free of AGN contamination.

3.0.3 The Far-Infrared Radio Correlation

In this section, I turn to the FIRC and how Shark can model the its relationships with other parameters. A benefit of any galaxy simulation is that a wide array of galaxy properties is tracked across the population. This makes Shark an ideal testing ground for FIRC and our understanding of the physics behind it. Previous Figures have shown that Shark reproduces FIRC for SFGs (Figure 3.1) and (U)LIRGs (Figure 3.3). In this section I discuss the FIRC in

terms of q_{IR} as defined in Equation ??.

3.0.4 Evolution of q_{IR} with M_*

In a recent paper, Delvecchio et al. (2021) found that rather than redshift, q_{IR} evolves with M_* . I begin my analysis of q_{IR} by examining whether this is true in Shark as well.

Figure 3.7 shows the evolution of q_{IR} with M_* with the number counts of galaxies. All galaxies from Shark with $10^8 M_{\odot} \le M_* \le M^{12} M_{\odot}$ and $0 \le q_{IR} \le 3.3$ at z = 0. Only cells with more than 100 galaxies are displayed. The median line and $-\sigma$ lines are of the same population, but are also influenced by M_* bins with fewer than 100 galaxies (This is why the median line appears to extend beyond the coloured region).

There do exist galaxies outside of this range however I do not display them here. Too few galaxies above $10^{12} \rm M_{\odot}$ are present to be plotted (This is evident by the number of blank cells above $11.5 \rm M_{\odot}$). Galaxies below $10^8 \rm M_{\odot}$ do not necessarily have physical counterparts. Galaxies with $q_{\rm IR} > 3.3$ do exist, but again are too few to be plotted meaningfully here. There are a substantial amount of galaxies with $q_{\rm IR} < 0$, but they had unphysical SFRs (on the order of $10^{-5} \rm M_{\odot} \rm yr^{-1}$), so have been excluded accordingly. In any case, plotting with the entire $q_{\rm IR}$ range gives a similar median line seen here, so including or excluding these galaxies would not make a substantive difference.

For $M_* \ge 10^{10} M_{\odot}$ has only a single population centered around $q_{IR} \sim 2.7$. This is on the scale that agrees with most observational results of q_{IR} (ie Bell (2003) found $q_{IR} = 2.64 \pm 0.02$). It is evident that this population extends to $M_* \le 10^{10} M_{\odot}$ as well, but the median descends to as low as $q_{IR} \sim 1.5$ at $M_* \sim 10^{(8.5)}$. This is due to the influence of a second population that has a lower I_{IR} . This second population has a q_{IR} as low as ~ 0.0 , but increases with a constant slope to meet the original population at $M_* \sim 10^{9.5} M_{\odot}$. The lower q_{IR} population is something of an anomaly; as best I can tell such divergent populations have not been observed or predicted in previous works.

With this as the starting point, the forthcoming plots will investigate the physical properties of this population of galaxies and explain why it exists.

Is it Infrared or Radio that causes this dip in q_{IR} ?

By Equation ?? a decrease in q_{IR} as seen in Figure 3.7 can either be caused by a decrease in L_{IR} or an increase in L_{radio} . Figure 3.8 shows that this is because of a decrease in L_{IR} . Figure 3.8 shows q_{IR} against M_* coloured by median IR luminosities (left column) and median radio luminosities (right column) properties. These plots focus in on masses $8_{\circ} \leq \log_{10}(M_*) \leq 9$ since these are the masses where the divergence is strongest. The top row shows IR and radio luminosities, with no normalisation in W and W respectively. Broadly this shows the expected behaviour of increasing luminosity with M_* . It is also clear that the lower q_{IR} population has a lower L_{IR} at the same M_* . L_{rad} would have to increase in order to produce the same effect on q_{IR} and it is not increasing. However, for completeness, I include the second rows for a

For that reason I normalise the luminosities to M_* in the second row. To do this, I divide the luminosities of each galaxy by its M_* , so that any evolution due to M_* is factored out. A clear picture forms for L_{IR} in this plot: for the higher, more constant q_{IR} , LIR/M_* is constant and on the high end of the plot. However for the lower, but increasing population of q_{IR} , LIR/M_* increases with increasing q_{IR} .

The normalised $L_{\rm rad}$ plot, on the other hand, is similar, but not quite as clear. The higher, constant population does have a high normalised $L_{\rm rad}/M_*$ and the lower, increasing $q_{\rm IR}$ population see a similar increase in $L_{\rm rad}/M_*$ to that seen in the $L_{\rm IR}/M_*$, but it is not as clear cut. These trends are general and there is a significant amount of scatter in these results. Further, the different scales used in $L_{\rm IR}$ and $L_{\rm rad}$ mean that a one-to-one comparison is not useful. The extent to which the luminosities change may be exaggerated if the limits of the colour scale are too small.

It is for that reason that in the final row of Figure 3.8, I plot ΔL_{IR} and ΔL_{rad} . ΔL_{IR} is defined as the distance from the median of the L_{IR}/M_* . The full definition is given in Equation 3.2:

$$\Delta L_{IR} = L_{IR}/M_*/[W/M_{\odot}] - median(L_{IR}/M_*/[W/M_{\odot}])$$
(3.2)

This puts the differences in L_{IR} and L_{rad} on the same scale. The colour scale in the final row is deliberately set to [-1,1] to highlight the amount of difference between the two luminosities. What this confirms is the significance that L_{IR} has on q_{IR} . The ΔL_{IR} extends over the entire [-1,1] range and increases with increasing q_{IR} . ΔL_{rad} , on the other hand, does increase with increasing q_{IR} , but its change does not extend over the entire range, instead a limited portion of it.

While the decreasing L_{rad} is not as impactful on q_{IR} than the decreasing L_{IR} (for the simple reason that L_{IR} decreases more), it is still important to note. The decreasing L_{rad} complicates the analysis as individual processes may influence L_{rad} or L_{IR} or perhaps both.

Overall, what can be concluded from this plot is that the decreasing q_{IR} is defined by a lower than average L_{IR} . In the next section, I will examine some of the properties of these galaxies to try and elucidate the physical characteristics of these lower q_{IR} galaxies.

The Physical Characteristics of lower q_{IR} galaxies.

Figure 3.9 shows the same population of Shark galaxies as in Figure 3.8, but coloured by different properties. These properties have been chosen because they either clearly indicate some diverging properties between the two q_{IR} populations or in previous papers a correlation between the property and q_{IR} has been found (or in the case of T_{eff} , both).

The upper left plot is coloured by effective dust temperature ($T_{\rm eff}$). In Shark, $T_{\rm eff}$ is not calculated directly, instead it is approximated using Equation 3.3

$$T_{eff} = T_{BC} \times L_{IR,BC} + T_{diff} \times (1 - L_{IR,BC}) \tag{3.3}$$

Where $T_{BC} = 50 K$ is the temperature of the birth clouds, $T_{\rm diff} = 25 K$ is the temperature of the diffuse ISM and $L_{\rm IR,BC}$ is the total IR emission from birth clouds as calculated from the formalism in Lagos et al. (2019). The temperatures of the birth clouds and diffuse dust have been set somewhat arbitrarily and so this is only an approximation. Modelled in this way $T_{\rm eff}$ acts as a measure of the strength of the contribution from birth clouds on the IR spectrum.

The T_{eff} plot then shows that high, constant q_{IR} is cooler and lower, increasing q_{IR} population is much hotter. Hence, the population of high, constant q_{IR} is dominated by emission from the cold ISM, whereas the emission from the lower, increasing q_{IR} is dominated by emission from hot birth clouds.

Smith et al. (2014) and Ivison et al. (2010a) also found a similar negative correlation between T_{eff} and q_{IR}. In particular, Smith et al. (2014) found that this relationship strengthens at longer IR wavelengths. The explanation that was given is since blackbody radiation has a strong dependence on wavelength, the smaller, hotter birth clouds can overcome the larger, cooler ISM. But these hot wavelengths are shorter than at hotter temperatures so will fade faster than the longer, cooler wavelengths.

Shark agrees with this finding. When a galaxy's IR emission is dominated by hot birth clouds, the integrated IR emission will have a greater contribution from shorter wavelengths. However, since these wavelengths fade faster than longer wavelengths, the total IR emission is reduced. This in turn reduces the IR as seen in Figure 3.8 and the overall reduction in q_{IR} seen in this section.

As M_* increases though, the proportion of birth clouds to disks shrinks. The birth clouds struggle to dominate over the increasingly massive ISM, which starts to contribute a greater proportion of the IR emission. This is why as M_* increases, the lower q_{IR} also increases until the emission from the ISM completely dominates the emission from birth clouds.

This explains the reason why the T_{eff} looks as it does.

The upper right plot is coloured by the normalised total gas mass (M_{gas}) . This is defined as by Equation 3.4:

$$M_{gas} = \frac{M_{gas,bulge} + M_{gas,disk}}{M_{\odot}} \tag{3.4}$$

where $M_{gas,bulge}$ and $M_{gas,disk}$ is the mass of the gas in the bulge and disk respectively, both in M_{\odot} .

Total gas mass has a strong dependence on M_* , so to remove any contribution from this evolution and instead focus on any evolution with q_{IR} , I have chosen to normalise it against M_* .

Both $M_{gas,bulge}$ and $M_{gas,disk}$ are fundamental baryonic quantities that Shark models directly. They are both tracked for galaxy and are easily accessible at the end of each snapshot (See Equations 49-64 in Lagos et al. (2018)). Both $M_{gas,bulge}$ and $M_{gas,disk}$ includes atomic gas and gas locked up in metals.

The M_{gas} figure in Figure 3.9 shows a complex picture. As it appears, particularly at $M_* \sim 10^8 - 10^{8.4}$, a higher q_{IR} can imply a higher M_{gas} . As best I can tell, there are no other studies

that have compared q_{IR} with M_{gas} , so the analysis here is purely based on Shark.

However, I contend that this matches the same picture as seen with $T_{\rm eff}$. Within the same M_* bins, a higher $M_{\rm gas}$ means that the gas to M_* ratio is higher. Hence, the ISM to birth cloud ratio is also higher. Thus, emission from the ISM will more easily dominate over the IR emission, making it more regulated.

The middle left plot is coloured by the half-mass radius of the disk ($r_{0.5mass}$). This is an indicator of the total size of the galaxy. Like $M_{gas,bulge}$ and $M_{gas,disk}$, $r_{0.5mass}$ is directly tracked in Shark, so it is simple to access them at the end of each snapshot.

What this shows is that within the same M_* bin, higher q_{IR} will have a smaller $r_{0.5 mass}$ meaning that they are smaller than galaxies with lower q_{IR} . Like M_{gas} this is tracked directly in Shark, so is easily accessed at the end of each snapshot.

Being bigger in size at the same M_{*} makes a galaxy more dispersed.

The middle right panel is coloured by gas surface density ($\Sigma_{\rm gas}$). The gas surface density is found via Equation 3.5.

$$\Sigma_{gas} = \frac{M_{gas}}{2\pi r_{0.5mass}^2} \left[\frac{M_{\odot}}{Mpc^2} \right]$$
 (3.5)

What this panel shows is evolution of q_{IR} with Σ_{gas} . Lower Σ_{gas} is associated with lower q_{IR} . The lower left plot is coloured by specific star-formation rate (sSFR). sSFR is defined in the usually way being SFR/M_{*}. As SHARK models L_{sync} as directly proportional to SFR and at 1.4GHz synchrotron radiation dominates over free-free emission, this makes SFR an indicator of radio emission. SFR is known to evolve with M_{*} however, so sSFR is used here to reduce the effect of this evolution.

What the sSFR plot shows is a negative correlation between sSFR and q_{IR} . This gives some explanation as to the decreasing L_{rad} seen in Figure 3.8. A decreasing sSFR means that there are less CCSNe to accelerate CRs to synchrotron wavelengths.

The lower right plot is coloured by SFR surface density (Σ_{SFR}). Σ_{SFR} is defined in a similar way to Σ_{gas} :

$$\Sigma_{gSFR} = \frac{SFR}{2\pi r_{0.5mass}^2} \left[\frac{M_{\odot}}{yrMpc^2} \right]$$
 (3.6)

This shows a weak correlation between Σ_{SFR} and I_{IR} , particularly at low M_* . Though it is important to note the strong correlation shown between Σ_{SFR} and M_* that appears to over-power this relation.

This is different to what was found in Delvecchio et al. (2021), which found an anti-correlation between q_{IR} and Σ_{SFR} . Seemingly driven by SFR, they see this relation independent of M_* . It is also in contrast to the models in Lacki et al. (2010), which predicted a anti-correlation as well, but steeper than what was found in Delvecchio et al. (2021).

This result is a combination of the result of sSFR and half-mass radius. Those plots showed an increased half-mass radius and an increase sSFR. It appears that there might be a conspiracy

whereby these two seem to cancel each other out when combined into $\Sigma_{\rm SFR}$.

Taken together, the individual plots of Figure 3.9 show that those galaxies with lower $_{\rm IR}$ in Shark have: (i) high $T_{\rm eff}$ and are thereby dominated by birth cloud emission, (ii) less gas, (iii) are larger and more disperse, (iv) lower $\Sigma_{\rm gas}$, (v) lower sSFR and (vi) lower $\Sigma_{\rm SFR}$. These properties will be brought together in Section 4 where I bring together the consequences of these results on our understanding of galaxy formation.

Figure 3.10 shows the IRX- β plane for Shark galaxies coloured by their q_{IR} value. Only cells with more than 100 galaxies within are shown.

The IRX- β relation is used as a measure of dust attenuation (Reddy et al. 2018). It is widely thought that the UV radiation emitted by young, massive stars is absorbed by the surrounding dust and re-emitted as IR radiation. The amount of excess IR (IRX) (as defined in Equation ??) represents how much UV radiation is absorbed and re-emitted in this process. A high IRX means more UV absorption and hence a high UV optical depth and a lower IRX means less UV absorption and a lower UV optical depth

$$IRX = \frac{L_{IR}}{L_{UV}} \tag{3.7}$$

 β is the UV spectral slope and is a measure of how the intensity of light changes with λ .

$$f_{\lambda} \propto \lambda^{\beta}$$
 (3.8)

where f_{λ} is the observed flux (Calzetti et al. 1994). β is often used as a proxy for the amount of dust obscuration (Reddy et al. 2018).

As discussed in the method section, Lagos et al. (2019) provides the UV-IR SED for all Shark galaxies. The luminosities required here are found easily from these results. The L_{IR} is the total luminosity integrated over $8-100\mu m$ (As it is through out this paper). L_{UV} used in the IRX calculation is at 1500Å. To calculate the β of each galaxy, the L_{UV} at 1500Å and 2500Å is calculated. The ratio of these fluxes and the wavelengths used readily gives β for each Shark galaxy.

The galaxies shown in the hexbin of 3.10 is a subset of all galaxies from Shark. The parameters for selecting this subset is; (i) all galaxies with a gas metallicity $(Z_{gas}) > 0$ $((Z_{gas})$ calculated following Equation 3.9) are included.

$$Z_{gas} = \frac{(M_{Z,disk} + M_{Z,bulge})}{(M_{gas,disk} + M_{gas,bulge})}$$
(3.9)

Where $M_{Z,disk}$ and $M_{Z,bulge}$ are the mass of metals in the disk and the bulge respectively. Both properties are tracked by Shark from snapshot-to-snapshot. That being said, there will be some galaxies that evolve without any metals in them. Such galaxies are unphysical and so are excluded.

(ii) Only central galaxies as defined in Shark are included. The definition of a central galaxy is involves the process by which Shark evolves galaxies. At z = 0, all halos and subhalos are

found using the halo finder code VELOCIRAPTOR. The biggest subhalo within each halo is defined as the central subhalo. This process is completed iteratively back through time. The central galaxy is then the galaxy that evolves within the central subhalo. Within Shark there are also satellite and orphan galaxies, the details of which are not important to this work, but further information is available in section 4.1 of Lagos et al. (2018).

(iii) A selection of galaxies based on their absolute magnitudes (AB) is made. Galaxies with an AB with dust between $-30 \le AB \le -10$ are included. This selection is made to exclude very bright and very faint galaxies that might skew the sample. The AB is taken from the results of Lagos et al. (2019). It is also to simulate a model of SF galaxies.

This is compared with the IRX- β relation found in Reddy et al. (2018). Reddy et al. (2018) found the IRX- β using obsevations of 3,545 SF galaxies over z = 1.5 – 2.5. To do so they assumed a three different dust attenuation curves, one of which is from Reddy et al. (2015), which is the one I use here.

That this curve follows the overall trend of the IRX- β relation found in Shark is more of a confirmation of the results from Lagos et al. (2019) than this work. Instead, the colours by q_{IR} is what is significant here.

Figure 3.10 shows a strong relationship between IRX and q_{IR} . One might expect this by considering equations 1.1 and 3.7, however given observational results suggest that q_{IR} is constant, it is surprising that this occurs. What this relationship suggests is that q_{IR} has a strong relationship to dust attenuation and that less dust attenuation leads to a lower q_{IR} . That is, galaxies with lower q_{IR} have less of the UV emission absorbed by dust and re-emitted into IR wavelengths.

Comparisons with the GAMA Survey

Figure 3.1 shows q_{IR} against stellar mass for SFGs. These are the same galaxies from SHARK (All SFGs at z=0) and GAMA that are shown in Figure 3.1. Again, SHARK is shown in blue, with the thick blue line showing the median of the population and the shaded blue region show $1-\sigma$ uncertainty. The black circles shows results form GAMA.

The red line shows the Equation 5 from Delvecchio et al. (2021):

$$q_{IR}(M_*, z) = (2.646 \pm 0.024) \times A^{-0.023 \pm 0.008} - B \times (0.148 \pm 0.013)$$
 (3.10)

where A = (1 + z) and B = $\log_{10} \left(\frac{M_*}{M_{\odot}} - 10 \right)$, where I have set z = 0. The red shaded region shows the uncertainty that comes from this equation.

Delvecchio et al. (2021) uses IR data from the Cosmic Evolution Survey (COSMOS) using de-blended far-IR/sub mm data from Jin et al. (2018). This is combined with radio data VLA-COSMOC 3GHz Large Project from Smolčić et al. (2017). For calculating q_{IR} , this 3GHz data is converted to 1.4GHz assuming a spectral index of $\alpha = -0.75$. Stellar masses and redshifts are determined using the optical-MIR photometry stellar synthesis library from Bruzual &

Charlot (2003), the results being the median from the liklihood distribution. The galaxies selected here have a z 0-4.5 and a stellar mass of $10^8 \le M_* \le 10^{12}$.

AGNs are selected and removed using a recursive methodology. After correcting for mass and redshift biases within the data, they calculate $q_{IR,peak}$ which represents the mode of the radio detections. A histogram is created about the $q_{IR,peak}$ which has a main peak to the right and a secondary peak in the left (See Figure 10 in Delvecchio et al. (2021). It is assumed that the main peak is due to SFGs and the secondary due to AGNs. Two Gaussians are then fitted about these two peaks representing the SFG and AGN populations respectively. This is done first by assuming the SFG histogram is symmetric, so they mirror the SFG histogram about its peak from right to left. All galaxies below 2σ of the SFG histogram are classified as AGNs and are removed. This corresponds to on average,0.43 dex below q_{peak} .

This process is then repeated recursively until such time as the $q_{IR,peak}$ does not change between fits. As it happens, they find it is only necessary to run this process twice.

The recursive AGN selection method is conducted only on galaxies with $M_* > 10^{10.5}$. The bins of galaxies below this are largely incomplete and therefore do not represent a representative sample of the stellar masses. Using their two complete bins they find a $q_{IR} - z$ relationship of $q_{IR} \propto (1+z)^{-0.055\pm0.018}$. Assuming that the remaining bins have the same slope of the $q_{IR} - z$ relationship, they extrapolate the recursive AGN removal method with the normalisation being left to vary. After this fit all galaxies below 0.43 dex of this best fit are identified as AGNs are removed.

As before, this process is repeated recursively until the median q_{IR} does not change between processes, outside of uncertainties.

Following this process, the authors fit a surface in the three-dimensional $rmq_{IR} - M_* - z$ space to find Equation 3.10. Two main findings of this paper was that q_{IR} does not evolve strongly with z, but does evolve with stellar mass.

Delvecchio et al. (2021) also compares with the results of Magnelli et al. (2015) and Delhaize et al. (2017). Following the same methods as outlined in these two papers (in particular AGN identification and removal), they find similar results with different data sets. This adds strong evidence that the tension between $q'_{IR}s$ evolution (or lack thereof) with redshift is due to different AGN identification procedures.

In the yellow line shows the result from Bell (2003) which is $q_{IR} = 2.64 \pm 0.02$. The shaded yellow region shows the uncertainty from this result. Bell (2003) uses IR data from the Infrared Astronomical Satellite (IRAS) and radio data from the NRAO VLA Sky Survey (NVSS). This uses data from local SFGs (z 0) and no stellar masses were estimated. AGNs have not been excluded from this sample, which makes it remarkable that the q_{IR} is so high. The result from Bell (2003) is cited in many papers as a standard result to which to compare.

Figure 3.11 allows for a four-way comparison of different q_{IR} results. From this figure, it is clear that the GAMA data, Bell (2003) data and SHARK data all agree in the $10^{10} - 10^{12} M_{\odot}$ range. As this is where the majority of the data from GAMA and Bell (2003) comes from, this is indicates that SHARK can reproduce the observations that lead to this data. There is some

scatter from the Shark data and the GAMA data, but that is to be expected.

There is very little GAMA data in mass ranges below $10^{10} - 10^{12} M_{\odot}$ and the data from Bell (2003) did not have stellar mass estimates, so comparisons here with SHARK are not meaningful. However, below $10^{10} M_{\odot}$, the SHARK q_{IR} drops off dramatically. The reason for this drop off, which will be discussed in depth in coming sections, is due to these low mass galaxies being no longer optically thick at UV wavelengths. In line with the calorimetric model, young massive stars produce UV radiation which is absorbed by dusty, molecular clouds in which these stars are formed and re-radiated in IR wavelengths. The key assumption here is that these clouds are totally optically thick to UV wavelengths. However for low mass galaxies, this assumption begins to fall away. The result is that these galaxies no longer produce the same amount of IR radiation due to star-formation.

Radio luminosities are unaffected by this; CCSNe will still accelerate cosmic rays into synchrotron radiation irrespective of the UV opacity of the galaxy.

It's for this reason why at these low mass ranges, Shark no longer agrees with Bell (2003). The striking feature of this Figure is the difference between the data from Shark and Delvecchio et al. (2021). At no stellar mass does the slope of Delvecchio et al. (2021) agree with Shark, though there is an intersection at $10^{9.5} \mathrm{M}_{\odot}$. There are many reasons for this, which I will now discuss.

Firstly, Delvecchio et al. (2021) assumes that the slope of the $q_{IR}-z$ relation is the same for high mass and low mass galaxies when removing AGNs. As I show in later, this is not the case in Shark. Shark predicts that the slope of $q_{IR}-z$ is positive, but flattens out for $10^8 \le M_* \le 10^9 M_{\odot}$ and this trend continues, but in a less dramatic way for $10^9 M_* \le 10^{10}$. Shark does predict a relatively constant slope for the $q_{IR}-z$ relation for SFGs with stellar mass $10^{10} M_* \le 10^{11}$ and $10^{11} M_* \le 10^{12}$. This lends credence to the logic behind the assumption Delvecchio et al. (2021) made. This was based on the results from the bins $10^{10.5} M_* \le 10^{11}$ and $10^{11} M_* \le 10^{12}$ in Delvecchio et al. (2021) which showed little slope evolution like Shark. Examining only these bins would lead to the assumption that the slope does not change.

The effect of this assumption is that many more SFGs are removed from the lower stellar mass bins in Delvecchio et al. (2021) than may be strictly necessary. If the observations were to be similar to the results from Shark, then many of these lower mass galaxies would have a lower q_{IR} . Later, I also show that the distribution of Shark galaxies in the $q_{IR} - M_*$ plane is biomodal. A relatively flat slope would more easily fit with the upper population of galaxies. Thus, when fitting the constant slope and removing all 0.43dex below this fit, there exists (at least in Shark) SFGs that would be removed in this process. While other papers can attribute their evolution with qir - z due to AGN identification process being not strict enough, perhaps the method used in Delvecchio et al. (2021) is too strict for low mass galaxies.

That being said, Shark may be over estimating the amount of SFGs at these masses. The proportion of galaxies which contain AGNs is high at low stellar masses (cite). So in that sense, the strength of these galaxies may be overstated.

The most credible explanation for while Delvecchio et al. (2021) (and indeed any other obser-

vational papers) do not predict that q_{IR} is so low at low stellar mass is because they would be difficult to detect. Low mass galaxies already have a lower luminosities and are therefore excluded from many flux limited surveys. Add to that that these galaxies are particularly IR faint, then the likelihood of a survey being able to detect a IR-faint, low mass galaxy is very slim. The radio component of such galaxies with also be faint, (though not as faint) so it would require the flux limits of two instruments being low enough to measure the radio and IR component of such a galaxy. This makes Shark's prediction particuarly difficult to confirm. A final point for why this difference exists is the possibility of new physics. Other papers such as Bell (2003) and Lacki et al. (2010) suggest conspiracies to explain FIRC. Such conspiracies are not included in Shark as the physics behind them are not widely accepted and there is significant difficulty in modelling them on the scale of a SAM. However, if such a conspiracy is the reason behind Delvecchio et al. (2021) observations, then Shark would not be able to reproduce the observational results that Delvecchio et al. (2021) found.

3.0.5 Evolution of q_{IR} with redshift

In this section, I weigh into the debate of evolution (or lack thereof) of q_{IR} with redshift. To do so, I consider the different populations of galaxies established thus far (all Shark galaxies, SFGs and (U)LIRGs) and measure their q_{IR} at different snapshots of Shark.

Figure 3.12 shows the evolution of q_{IR} with redshift for all Shark galaxies and SFGs for different stellar mass bins. This is compared with results from Mao et al. (2011) and Delvecchio et al. (2021).

In the $8.0 \le \log_{10}(M_*)9.0$ bin a relationship emerges with q_{IR} galaxies that is visually similar to the $q_{IR} - M_*$ evolution. Starting at about $q_{IR} \sim 2$, q_{IR} increases with redshift until it appears to asymptotically tend towards the result from Mao et al. (2011). This evolution is remarkable in that, compared with observational results that suggest that q_{IR} evolves with redshift, it is evolving, but in the wrong direction. This is seen in the significant disagreement that the results from Shark has with Delvecchio et al. (2021). This evolution is seen in SFGs as well with both lines being nearly indistinguiable from each other.

Firstly, why is this evolution seen? It's not because of the model developed in the previous section of a birth cloud dominated galaxy having lower q_{IR} because of lower IR, rather it's likely the radio that has increased. With increasing redshift, SFR increases (See Figure 3.6). This increased SFR will directly increase the L_{rad} that is emitted from this galaxy. This can counteract the decrease in L_{IR} that is caused by the birth cloud dominated galaxies.

Secondly, why the big discrepancy with Delvecchio et al. (2021)? As discussed in section 3.0.4, the result from Delvecchio et al. (2021) finds the q_{IR} – z slope for higher mass galaxies and extrapolates this to lower mass galaxies. This result from Shark questions the validity of this assumption. From other panels, it is possible to see that this is not the case for Shark galaxies. Assuming this slope leads to an over-zealous removal of lower mass galaxies with lower q_{IR} , mistaking them for AGNs. This means that the Delvecchio et al. (2021) will predict a higher $_{IR}$

than what is seen in Shark.

In the $9.0 \le \log_{10}(M_*)10.0$ and $9.0 \le \log_{10}(M_*)10.0$ panels show similar results, so I will discuss them both at the same time here. A similar relationship between q_{IR} and redshift is seen as in $8.0 \le \log_{10}(M_*)9.0$, but to a less extent. At this stellar mass, galaxies are no longer birth cloud dominated, and their IR emission starts to increase as a result. For both the whole and SFG sample a slight increase in q_{IR} is seen reflecting, as seen in lower masses, the increase in SFR with redshift. Overall, q_{IR} is higher than at lower masses, indicating that the birth clouds and ISM are well mixed so the IR emission does not drop. This stellar mass bin has the best agreement with observations.

The final panel shows the only decoupling of SFGs and whole galaxy sample. In the definition of a SFG used in Shark, the average SFR will be higher since all galaxies from -0.3 dex and above will be selected form the MS. In other mass bins, this has not made a material difference because galaxies of those masses do not vary from the galaxy as much MS, but in these mass bins, there is quite a bit of variance. Further, the MS fit is done using galaxies $9.0 \le \log_{10}(M_*)10.0$, so it would be unreasonable to expect much deviation for those galaxies or those with stellar masses near it. But $11.0 \le log_{10}(M_*) \le 12.0$ is two orders of magnitude away from the MS fit, so the MS fit starts to deviate from where the average galaxy lies on the SFR $-M_*$ plane. Having a higher SFR means that the SFGs will have a higher radio emission. A high radio emission means they will have a lower q_{IR} .

It's also at this stellar mass bin that Shark simply runs out of galaxies of this mass. Only bins with ten or more galaxies are plotted, so no meaningful relation can be seen for these large galaxies beyond $z \sim 4$.

Like the lowest mass bin, the results here disagree with the relation found in Delvecchio et al. (2021), but in the opposite direction; Shark now predicts a higher q_{IR} than Delvecchio et al. (2021). This is true for both SFGs and the entire sample, even accounting for the SFR difference between the two. It's possible that AGN contamination in Delvecchio et al. (2021) is causing this difference.

Overall, Figure 3.12 finds a novel result that what is found in the literature. Mainly that q_{IR} does evolve with redshift, but it increases rather than decreases. However, the main driver of q_{IR} evolution is stellar mass. To some extent this confirms the hypothesis from Delvecchio et al. (2021), but in the opposite direction.

Figure 3.13 shows the evolution of q_{IR} vs. z for the (U)LIRG population across different stellar mass bins. This is compared with the results from Lo Faro et al. (2015) and Sargent et al. (2010b) both of which found observationally q_{IR} for (U)LIRGs. A caveat to keep in mind is that Lo Faro et al. (2015) did not directly compare the stellar mass or redshift evolution of (U)LIRGs, so the average of the whole population is shown here. With so few galaxies there is no statistical meaning to separating the populations by stellar mass or redshift, so the median of the population is shown in each panel.

There is a similar caveat for Sargent et al. (2010b), except that they did compare q_{IR} against redshift, but not against stellar mass.

The results from the stellar mass bins $8.0 \le log_{10}(M_*) \le 9.0$, $9.0 \le log_{10}(M_*) \le 10.0$ and $11.0 \le log_{10}(M_*) \le 12.0$ are remarkably similar, so the analysis will deal with all three at once. The exception is that at $8.0 \le log_{10}(M_*) \le 9.0$, there are no ULIRGs that can be plotted. This is not surprising since luminosity directly scales with stellar mass, so its natural that highly luminous galaxies will not exist at the lower stellar masses in high quantities.

Overall, these plots are rather dull and agree more with Sargent et al. (2010b) than with Lo Faro et al. (2015). As discussed previously, Sargent et al. (2010b) contains many more galaxies than Lo Faro et al. (2015), so Sargent et al. (2010b) can be interpreted as being more robust. At these stellar masses, the differences between LIRGs and ULIRGs is hardly relevant. This is presents a conspiracy of sorts. By definition, ULIRGs will have a higher L_{IR} and a higher SFR. This higher SFR will manifest as a high L_{rad} , so cancels out the increase in L_{IR} when calculating q_{IR} . What is remarkable about the similarity between these two populations is that it implies that the increase in SFR directly proportional to L_{IR} . If it wasn't a much bigger difference between the populations would be seen.

This conspiracy appears to break down in the $11.0 \le log_{10}(M_*) \le 12.0$. Here ULIRGs have the same q_{IR} compared with other stellar mass bins, whereas the LIRGs have a higher q_{IR} . As seen in Figure 3.3, at this higher stellar mass, there is an increase in the SFR of the ULIRG population, which is not seen in the LIRGs. This increase in SFR appears to continue to balance the increased L_{IR} in ULIRGs to keep q_{IR} constant. However, for LIRGs, SFR drops so q_{IR} increases.

This effect is puzzling. ULIRGs have a finely balanced q_{IR} , but LIRGs not. The strong agreement with Sargent et al. (2010b) is explained by the fact that Sargent et al. (2010b) only considered ULIRGs, so measured this finely balanced mechanism too. It also agrees with the findings of Lo Faro et al. (2015), that LIRGs have a higher q_{IR} than ULIRGs, when M_* is disregarded. The impact on this result is to suggest that long held assumptions that $L_{IR} \propto SFR$ may be true only for $M_* \leq 10^{11} M_{\odot}$ or for $L_{IR} \geq 10^{12} L_{\odot}$.

3.0.6 Lightcone

In this section I introduce the radio lightcone. This converts the luminosities calculated for each galaxy into fluxes based on their redshift in order to model how these galaxies would appear in observations. From this I calculate the number counts of galaxies for seven different radio frequencies and compare this with observational data of number counts. I then show the number counts, median M_* and SFR evolution with redshift at seven different frequencies.

The short version how this is done is to calculate the luminosity distance of each galaxy by first finding the co-moving distance for each galaxy's redshift. For completeness I will include the cosmological derivation here, but it is done in the exact way one would expect.

To convert from the radio luminosities calculated in Section 2.1, I first find the luminosity distance, D_L per Equation 3.11.

$$D_L(z) = (1+z) \times D_M(z)$$
 (3.11)

where D_M is the co-moving distance. To find the co-moving distance I must first define the dimensionless Hubble parameter, E(z):

$$E(z) = \sqrt{\Omega_r (1+z)^4 + \Omega_m (1+z)^3 + \Omega_k (1+z)^2 + \Omega_\Lambda}$$
 (3.12)

where $\Omega_{\rm r}$, $\Omega_{\rm m}$, and Ω_{Λ} are the values of the radiation energy, matter and dark energy densities respectively. Ω_k determines the curvature of the Universe here being defined as $\Omega_k = 1 - \Omega_{\rm r} - \Omega_{\rm m} - \Omega_{\Lambda}$.

SHARK assumes the cosmology of a flat- Cold Dark Matter (CDM) model using the results from Planck Collaboration et al. (2016). Specifically, a flat-CDM model means that $\Omega_k = 1$ and the results of Planck Collaboration et al. (2016) give $H_0 = 67.51 \pm 0.64 \text{kms}^{-1} \text{Mpc}^{-1}$, $\Omega_m = 0.6879/\text{pm}0.0087$, $\Omega_{\Lambda} = 0.3121/\text{pm}0.0087$ and $\Omega_{\stackrel{\circ}{=}0}$.

The co-moving distance is then equal to the transverse comoving distance D_C , in a flat Λ . This is found using equation

$$D_C(z) = D_H \int_0^z \frac{z'}{E(z')}$$
 (3.13)

where D_H is the Hubble distance: $D_H = \frac{c}{H_0}$, where c is the speed of light.

For a given Ω_{Λ} the $D_{M}(z)$ can be easily calculated using the python package ASTROPY.

At this point I calculate the radio luminosity as per Section 2.1. However, one key difference using the rest frame frequency, $\nu_{\rm rest}$ found using Equation 3.14.

$$\nu_{\text{rest}} = \nu \times (1 + z) \tag{3.14}$$

where ν is the frequency that I am interested in. In this study I consider seven different frequencies: 150 MHz, 325 MHz, 610 MHz, 1400 MHz, 3000 MHz, 5000 MHz and 8400 MHz. $\nu_{\rm rest}$ is the frequency that the light will have after accounting for redshift due to expansion of the Universe. It's important to note that there's no observer that's modelled in Shark, so there's no line of sight or differences in motion of the objects that can be included. For that reason, only expansion of the Universe will affect the nu_{rest}, not relativistic redshift or the effects of gravitational potentials.

The flux of the galaxy is found by Equation 3.15

$$F = \frac{L_{\text{rad}}(\nu_{\text{rest}})}{4\pi D_{\text{L}}^2}$$
 (3.15)

where $L_{rad}(\nu_{rest})$ is the radio luminosity calculated at the rest frequency. Following this process, I determine a flux for every Shark galaxy and can determine the number count of galaxies at different frequencies and the physical properties of galaxies above a certain flux.

In this section I also make a distinction between the Bressan and Dale models. As discussed in section 2.1, the Dale model uses empirical dust templates to calculate a galaxies UV to IR SED and then converts that to radio frequencies by assuming a constant q_{IR} . This assumption makes it inappropriate to use the Dale model in the analysis in 3.0.3 (and as section 3.0.3 demonstrates, this assumption may not even be true). However, this assumption does not prohibit the Dale model's use in the lightcone. As such, both models are used here in this plot.

Number Counts

Figure 3.14 shows the radio source counts for the Bressan and Dale models. The radio source counts are found after the flux of each galaxy has been found for each of the seven frequencies, as per the formalism above. For each frequency a histogram of these fluxes is created. This histogram is then divided by the length of each bin (as an aside, this histogram is created in log-space so the length of each bin changes from bin to bin) times the effective survey area. In Shark, each subvolume is a box with side lengths 210 Mpc. This corresponds to an area of 1.67deg^2 , so the total area of that all 64 subvolumes cover is 108deg^2 .

This is compared with observational results from Tompkins et al. (2023). Tompkins et al. (2023) includes a comprehensive compendium of radio source data from a variety of papers, frequencies and instruments. This includes data from Huynh et al. (2015); ? (8400 MHz and 5000MHz) Butler et al. (2018); Gordon et al. (2021); Vernstrom et al. (2014, 2016); Van der Vlugt et al. (2021) (3000 MHz), Biggs & Ivison (2006); ?); ?); Huynh et al. (2005); Matthews et al. (2021); Prandoni et al. (2000); Seymour et al. (2008); ? (1400 MHz), Bondi et al. (2007); Garn et al. (2008); Hale et al. (2021); Ibar et al. (2009); Moss et al. (2007); Ocran et al. (2020) (610 MHz), Mazumder et al. (2020); Oort (1988); Riseley et al. (2016); Sirothia et al. (2009) (325 MHz) and Franzen et al. (2016); Hardcastle et al. (2021); Hurley-Walker et al. (2017); Mandal et al. (2021); Williams et al. (2016) (150 MHz).

In this Figure I also make a distinction between the Bressan and Dale models. As discussed in section 2.1, the Dale model uses empirical dust templates to calculate a galaxies UV to IR SED and then converts that to radio frequencies by assuming a constant q_{IR}. This assumption makes it inappropriate to use the Dale model in the analysis in 3.0.3 (and as section 3.0.3 demonstrates, this assumption may not even be true). However, this assumption does not prohibit the Dale model's use in the lightcone. As such, both models are used here in this plot. Due to the strong similarity between the two models, it is safe to assume that the Dale model would produce similar results in the upcoming Figures.

Note that Tompkins et al. (2023) includes a comparison with these results with "SHARK". This is the Dale model that is used here with an extension to include modelled AGN emission. Here I do not include AGN emission, but I do include the Bressan model.

Figure 3.14 displays the veracity of the Bressan and Dale models. These two different methods, which vary almost completely in their method of determining radio emission, produce almost identical results here. More than that these two models produce results that closely match the

results from the papers listed in Tompkins et al. (2023).

One worrying feature of these plots is that at every frequency, the models agree with the observational data before a point where the models appear to underpredict what is seen observationally (For example from $\log_{10}(S) > 1 \text{mJy}$ at 150 MHz), before stopping completely. The reason for this is that observations include fluxes from AGNs, whereas these models do not. The effect of these AGNs (likely radio loud AGNs) is to have sources which are at a much higher flux than the normal galaxies used here. High quantities of normal galaxies are not capable of producing such high fluxes at these frequencies, whereas radio loud AGNs are able to. The Shark model presented in Tompkins et al. (2023) did include a model for this AGN emission and matched the fluxes shown here. I discuss the possibility of adding this AGN model to the model used in this thesis in the Future Work section.

Number counts with z

Figure 3.15 shows the number of galaxies with a flux $S > 10^{-2} mJy$ per redshift. $S > 10^{-2} mJy$ was a limit that was chosen on the basis that it displays the overall trends the clearest, but there's no reason to expect that other limits would produce different results. This plot is calculated easily; having assigned each galaxy a flux at each frequency a histogram is created of the redshifts of each galaxy with a flux $S > 10^{-2} mJy$. This is then divided by the length of the bins used in the histogram and the effective survey area that has been modelled in Shark. The median is then plotted with the $1 - \sigma$ area approximated using bootstrapping.

The result shown here mimics that of a flux limited survey. The number of galaxies with $S > 10^{-2} \text{mJy}$ decreases with increasing redshift. At high z, the only galaxies that will have a high enough luminosity will be those that are very bright and very large.

This figure also displays the need for K-correction. K-correction is a process by which observed frequencies are shifted to the rest frequencies by considering the effect of redshift. With increasing redshift, the SED spectrum moves to lower frequencies as seen in Equation 3.14. For example, processes that emit at 325MHz at redshift = 1.166 will be observed at a frequency of 150MHz. Higher frequencies require more energy to produce, meaning that even at redshift 0, more objects are brighter at 150MHz than 8400 MHz. As redshift increases the frequencies required to produce a rest frequency of 8400 MHz gets higher and higher, and fewer and fewer galaxies are capable of producing enough of these high energy photons in order to overcome the flux limit imposed. This effect increases until such point as the amount of galaxies producing these high frequencies is no longer statistically meaningful; Shark runs out of galaxies that are able to produce these high frequencies.

This plot thus probes the extremes of Shark and the Bressan model.

Figure 3.16 shows the median redshift of galaxies above the flux threshold ($S_{threshold}$). This is found by determining the redshift distribution of all galaxies above $S_{threshold}$ and computing the median and $1 - \sigma$ percentiles of this distribution.

This figure is quite similar to Figure 3.15 in that it mimics a flux limited survey. For all

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frequencies it shows that the highest redshift is achieved with the lowest $S_{threshold}$. As the $S_{threshold}$ is increased, so the median redshift decreases. This reflects that very far galaxies will have a lower flux and so will not reach the $S_{threshold}$, thus will not be included in calculating the median.

Notably the 8400, 5000 and 3000 MHz lines all stop at a $S_{threshold} \sim 1 \text{mJy}$, indicating that beyond that too few galaxies are present to make a meaningful trend.

Like Figure 3.15, this figure too shows the need for K-correction. This is shown by the stratification of the frequencies from high to low. The 150MHz can rely on the higher frequencies to support it in that processes at high redshift will end up being observed at 150 MHz in the rest frequency. 8400 MHz on the otherhand is left out to dry. Only extreme processes at high redshift will be able to produce 8400 MHz in the rest frame and given the rarity of such processes it's not surprising that the median redshift of 8400 MHz is lower than all other frequencies at all $S_{\rm threshold}$

Stellar Mass and SFR with z

In the final figure of this section, Figure 3.17 shows the distribution of M_* and SFR with redshift for galaxies with $S > 10^{-2}$ mJy with the upper panel showing M_* and the lower, SFR. The median of these quantities are found in the exact same way to those in Figure 3.15.

Both panels show remarkably similar behaviour. At low redshifts, the M_* and SFR is low, but this increases asymptotically with redshift. This again mimics the behaviour of a a flux limited survey. Only high mass galaxies are capable of producing enough light to be seen at high redshifts. Similarly, in this model where SFR is proportional to $L_{\rm rad}$, only highly star-forming galaxies can be seen at high redshifts.

This acts as an indication of the biases that high redshift surveys have to account for. In flux limited surveys, only massive and highly star-forming galaxies are able to be observed. This will skew results to the properties of these galaxies. This conclusion aligns with the thesis of Delvecchio et al. (2021) which was that evolution of q_{IR} is due to stellar mass and that any evolution seen with redshift would be due to this bias.

As in other plots, there is a stratification of frequencies, except in the opposite way. Higher frequencies require higher M_{*} and SFR than lower frequencies at the same redshift. Again, this is indicative of the need for K-correction. Only very massive galaxies are capable of producing the high energy photons required to have a high enough frequency to be able to be observed at a rest frequency of 8400 MHz.

Old qir plots

qir vs. z

3.0.7 Number Counts

Lightcone

Number counts with z

Stellar Mass and SFR with z

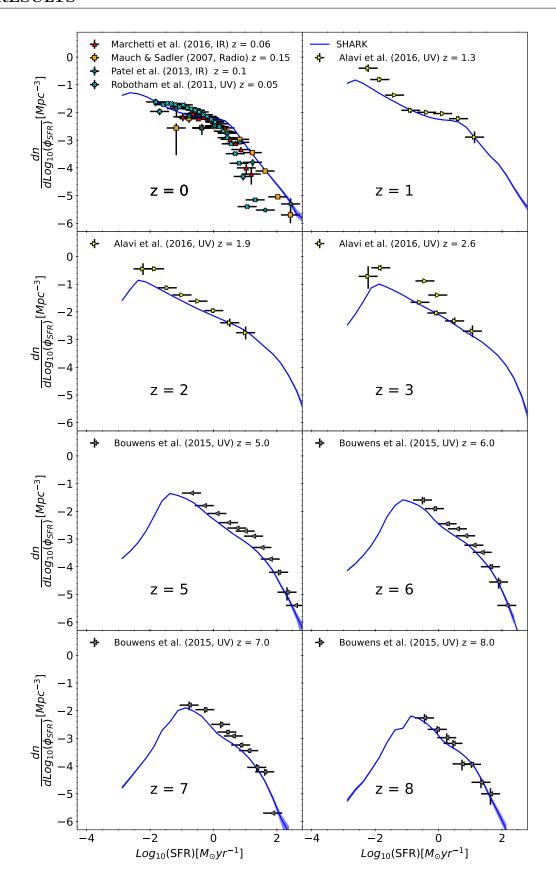


Figure 3.6: The SFR function over different redshift bins compared with observational data from Marchetti et al. (2016), Mauch & Sadler (2007), Patel et al. (2013), Robotham et al. (2011), Alavi et al. (2016) and Bouwens et al. (2015). as converted to SFR densities in Katsianis et al. (2017). The results listed in Appendix A of Katsianis et al. (2017) is plotted here. Like other plots, SHARK is shown in blue and the area represents the estimate of the $1-\sigma$ from bootstrapping.

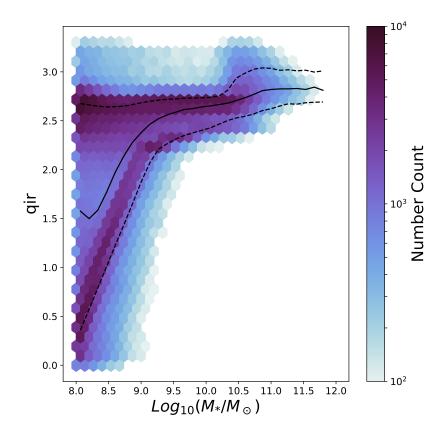


Figure 3.7: q_{IR} against M_* with coloured by number counts. The solid black line represents the median of the population and the dashed lines are $1 - \sigma$ uncertainty. Shown here are all Shark galaxies with $10^8 M_{\odot} \leq M_* \leq M^{12} M_{\odot}$ and $0 \leq q_{IR} \leq 3.3$ at z = 0. Only cells with more than 100 galaxies are displayed. Above $10^{10} M_{\odot}$, there is a single population of galaxies centered around q_{IR} 2.7 with no evolution evident. Below this there's evidence of two populations: one centered around q_{IR} 2.7 and a second that diverges from this population to a lower q_{IR} 0.5. There is evidence of evolution with M_* , but in the opposite way that Delvecchio et al. (2021) found. The colour map is from the package CMOCEAN (Thyng et al. 2016).

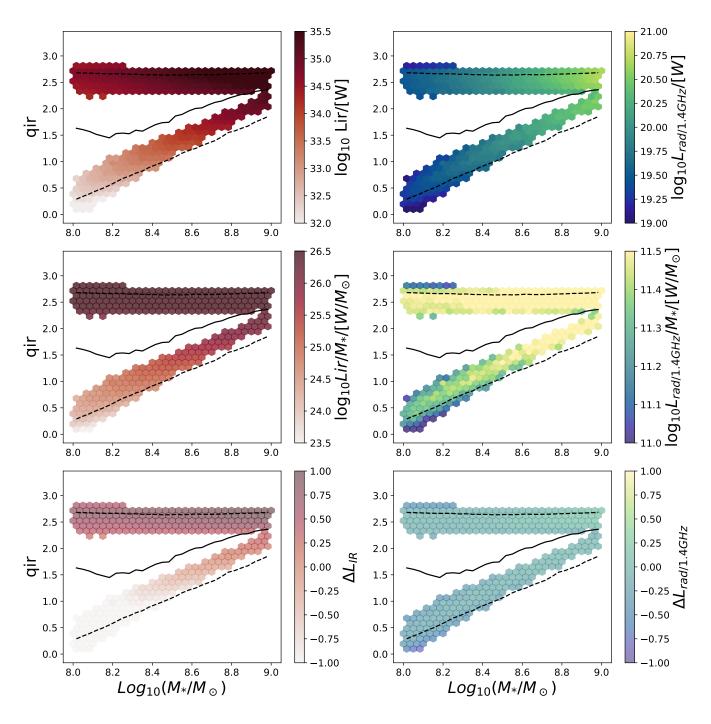


Figure 3.8: q_{IR} against M_* coloured by median IR luminosities (left column) and median radio luminosities (right column) properties. These plots focus in on masses $8_{\odot} \leq \log_{10}(M_*) \leq 9$ at z = 0. Only cells with 500 galaxies are shown here. As in Figure 3.7, the solid black line is the median of the population and the dashed lines represent $1 - \sigma$ uncertainty. The top row shows actual IR and radio luminosities in W and W respectively. The second shows IR/M* and $Radio/M_*$ to remove any evolution the luminosities can have with M_* . The bottom plots show ΔL_{IR} and $\Delta L_{rad/1.4GHz}$ where this is the difference between the population and the median luminosity/M*. This Figure shows that while both L_{IR} and $L_{rad,1.4GHz}$ decrease in this second population, L_{IR} decreases more leading to the decline in q_{IR} . The colour maps are from the package CMOCEAN (Thyng et al. 2016).

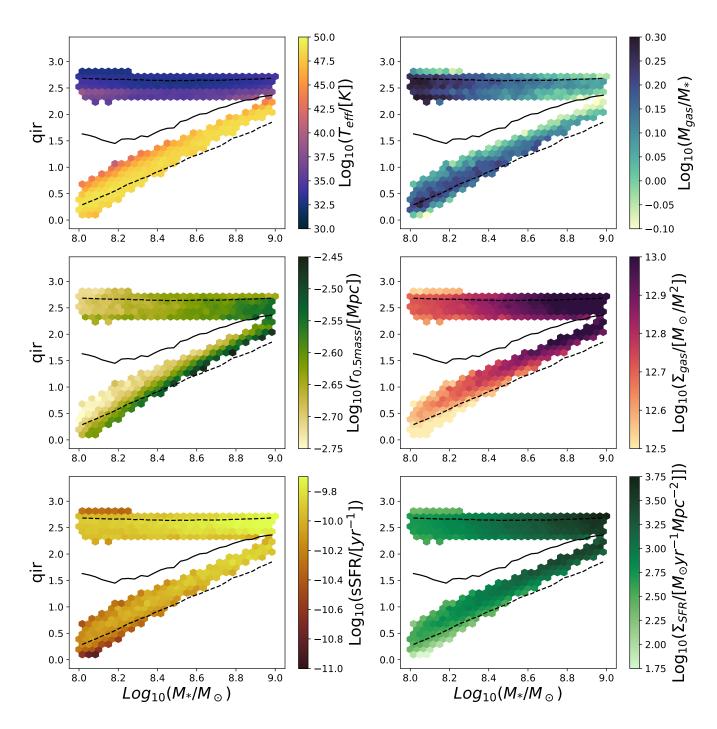


Figure 3.9: q_{IR} against M_* coloured a variety of properties. These plots focus in on masses $8_{\circ} \leq \log_{10}(\mathrm{M_*}) \leq 9$ at z=0. As in Figure 3.7, the solid black line is the median of the population and the dashed lines represent $1-\sigma$ uncertainty. Only cells with 500 galaxies are shown here. The upper left plot is coloured by effective dust temperature ($\mathrm{T_{eff}}$). The upper right plot is coloured by the total gas mass ($\mathrm{M_{gas}}$). The middle left plot is coloured by the half-mass radius of the disk ($\mathrm{r_{0.5mass}}$). The middle right plot is coloured by gas surface density ($\mathrm{\Sigma_{gas}}$). The lower left plot is coloured by specific star-formation rate (sSFR). The lower right plot is coloured by SFR surface density ($\mathrm{\Sigma_{SFR}}$). The colour maps are from the package CMOCEAN (Thyng et al. 2016).

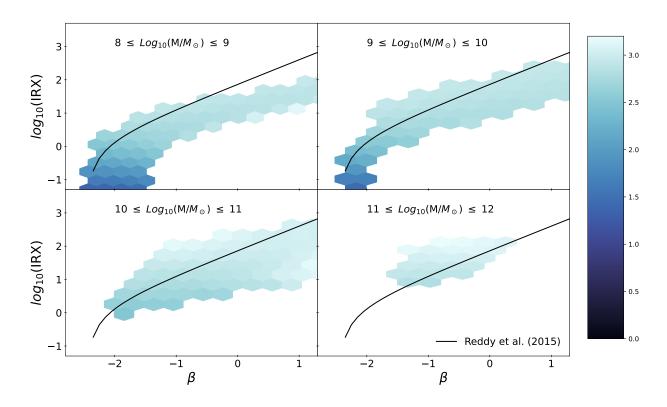


Figure 3.10: The IRX- β plane for different stellar masses coloured by the median q_{IR} of the bin. A subset of galaxies from Shark are selected for this plot (See text) at z = 0. This is compared with the IRX- β plane as found using dust attenuation curves found in Reddy et al. (2015). Only cells with more than 100 galaxies within are shown. The colour map is from the package CMOCEAN (Thyng et al. 2016).

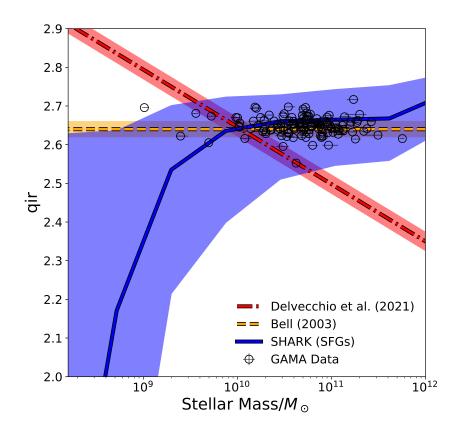


Figure 3.11: qir vs. total stellar mass compared with observational data of individual galaxies from the GAMA survey. These are the same galaxies as in Fig. ??. The blue region shows galaxies from Shark, with the thick line being the median result and the shaded blue region showing $1 - \sigma$ uncertainty. The red dot-dashed line shows the qir vs. total stellar mass relationship found in Delvecchio et al. (2021) (Fig. 14 in that paper). The orange dashed line shows the constant qir found in Bell (2003).

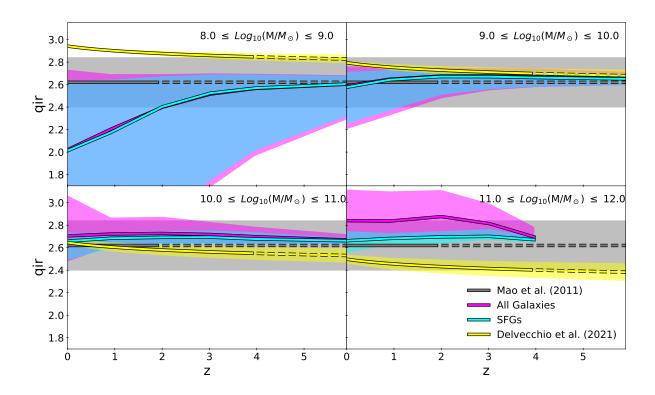


Figure 3.12: q_{IR} vs. z for different stellar mass bins for all galaxies in Shark (shown in magenta) and those classified as SFGs (shown in cyan). As before, the thick magenta/cyan lines represent the median of that particular redshift bin and the shaded area represents $1 - \sigma$ uncertainty. The grey line represents the result from Mao et al. (2011) (2.620.22) with the grey shaded area representing the uncertainty found in that relation. Mao et al. (2011) only observed out to z = 2, so after this redshift, this result is a projection, which is indicated by the dashed line. They yellow line shows the result from Delvecchio et al. (2021) as defined in Equaiton 3.10. Delvecchio et al. (2021) only included observations to z = 4.5, so after this point the relation is extrapolated which is indicated using a dashed line.

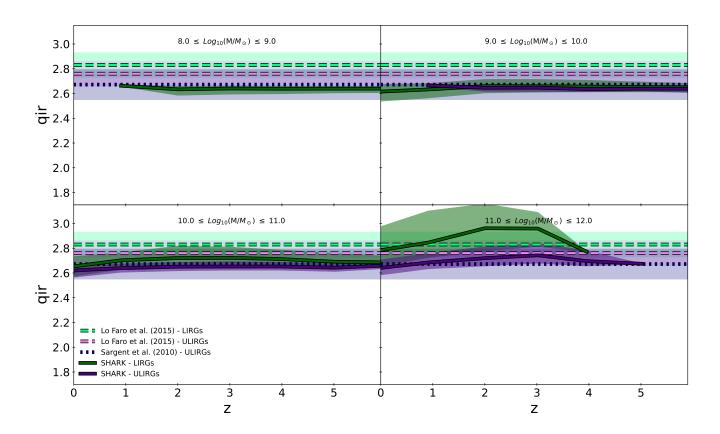


Figure 3.13: q_{IR} vs. z for different stellar mass bins for (U)LIRGs where LIRGs are shown in green and ULIRGs are shown in purple. As before, the thick line shows the median and the shaded area shows the $1 - \sigma$ uncertainty. This is compared with results from Lo Faro et al. (2015) and Sargent et al. (2010b). The galaxies used here are the same as the as in Figure 3.3.

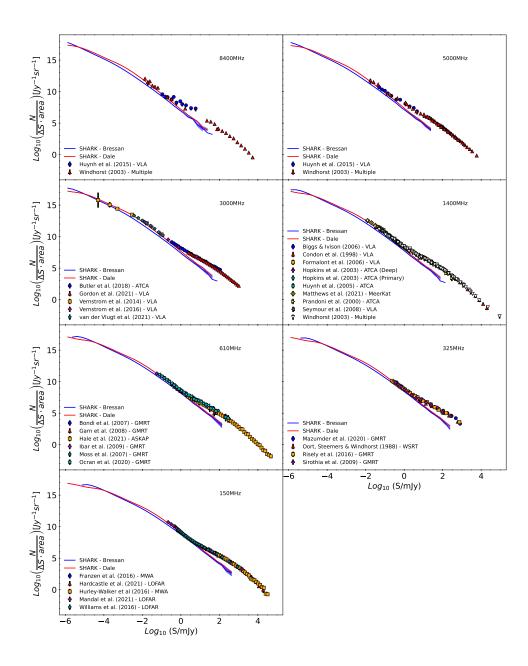


Figure 3.14: Radio source counts for the frequencies (from left to right): 8400 MHz, 5000 Mhz, 3000 MHz, 1400 MHz, 610 MHz, 325 MHz and 150 MHz. Galaxies modelled in Shark are shown in the two lines, the Bressan model in blue and the Dale model in red. Data points show counts from different observations from a variety of sources as compiled in Tompkins et al. (2023). See the text for the full reference list. List next to each reference is the telescope used to find such results.

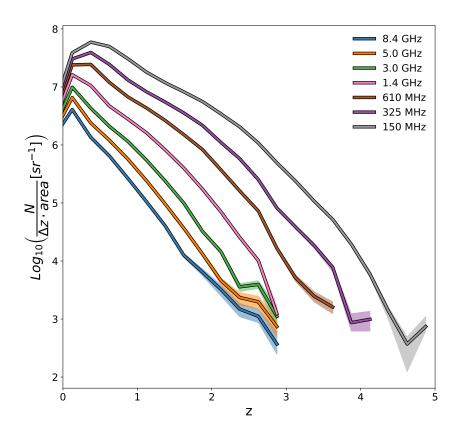


Figure 3.15: The median radio source counts per redshift for different radio frequencies for all galaxies with S > 10^{-2} mJy. Solid lines are the median line and the shaded area represents $1-\sigma$ estimated using bootstrapping. The blue line is 8400 MHz, the orange 5000 MHz, green 3000 MHz, pink 1400 MHz, brown 610 MHz, purple 325 MHz and grey 150 MHz.

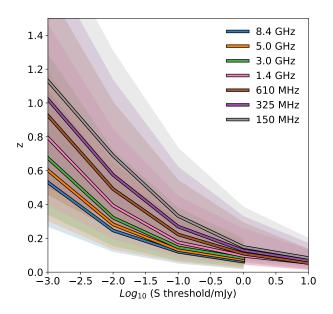


Figure 3.16: The median redshift of galaxies above a certain $S_{threshold}$. Solid lines are the median line and the shaded area the $1-\sigma$ percentile range of the population. The blue line is 8400 MHz, the orange 5000 MHz, green 3000 MHz, pink 1400 MHz, brown 610 MHz, purple 325 MHz and grey 150 MHz.

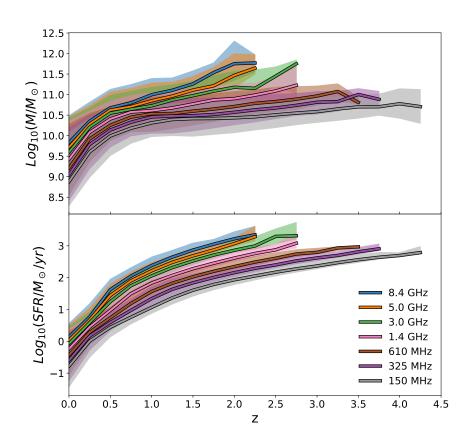


Figure 3.17: Upper panel shows the median total stellar mass of all galaxies with S > 10^{-2} mJy for different radio frequencies. Shaded area represents $1-\sigma$ region. Lower panel shows the same but median SFR. The blue line is 8400 MHz, the orange 5000 MHz, green 3000 MHz, pink 1400 MHz, brown 610 MHz, purple 325 MHz and grey 150 MHz.

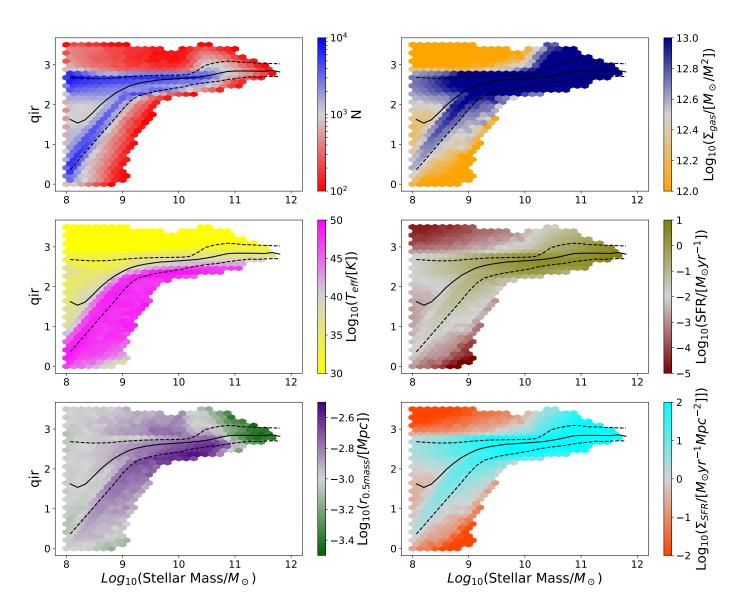


Figure 3.18: qir vs. total stellar mass coloured by different parameters. Upper left most panel shows number counts. Upper right panel shows coloured by effective dust temperature. Lower left panel shows coloured by half mass radius and lower right panel shows coloured by gas surface density. Solid black line shows the median qir evolution with total stellar mass and the dashed black lines show $1 - \sigma$ uncertainty.

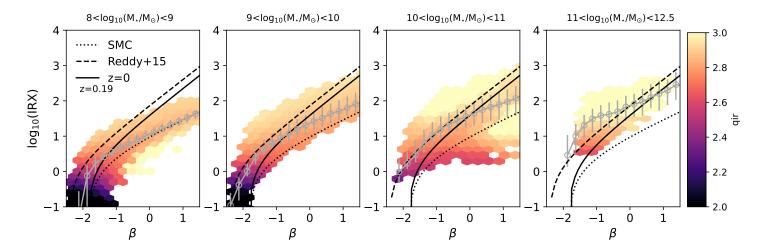


Figure 3.19: IRX-Beta function plot at z = 0.19 coloured by qir.

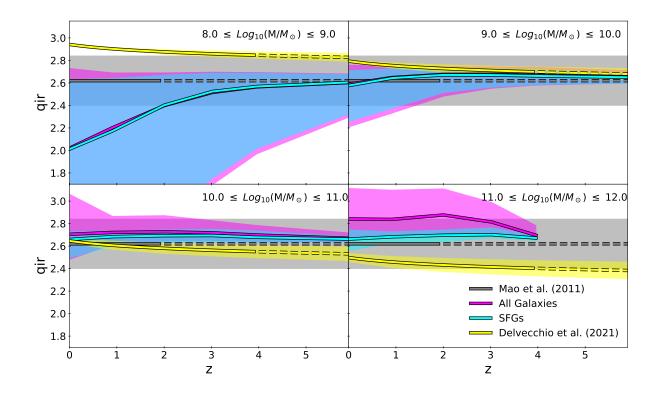


Figure 3.20: qir vs. z for different stellar mass bins. Solid line shows the median and the shaded regions show the $1-\sigma$ uncertainty.

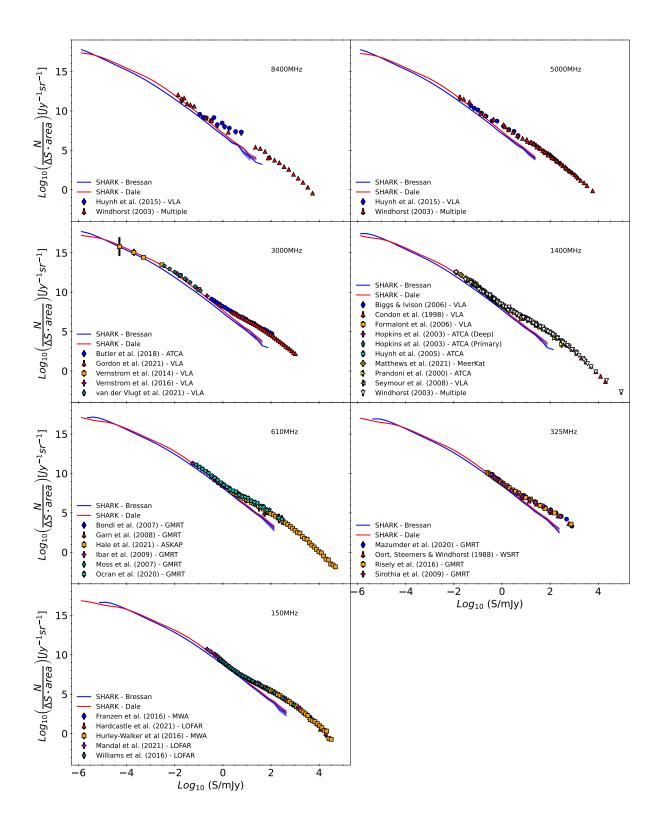


Figure 3.21: The number counts per redshift for different radio frequencies for all galaxies with $/rmS > 10^{-2}mJy$. Solid lines are the median line and the shaded area represents $1 - \sigma$ estimated using bootstrapping.

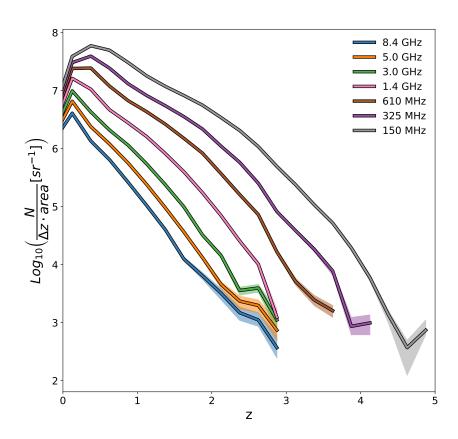


Figure 3.22: The number counts per redshift for different radio frequencies for all galaxies with $S > 10^{-2} mJy$. Solid lines are the median line and the shaded area represents $1 - \sigma$ estimated using bootstrapping.

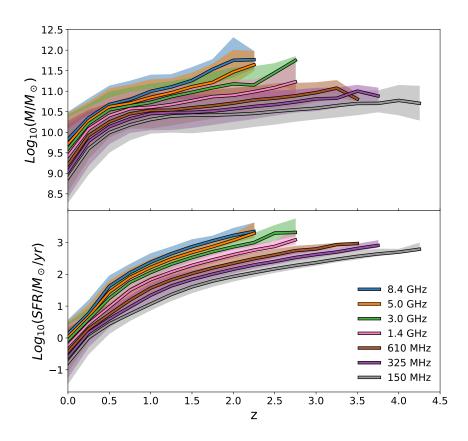


Figure 3.23: Upper panel shows the median total stellar mass of all galaxies with S > 10^{-2} mJy for different radio frequencies. Shaded area represents $1 - \sigma$ region. Lower panel shows the same but median SFR.

Chapter 4

Discussion

In this thesis I have introduced a radio galaxy emission model into the Shark semi-analytic model of galaxy formation. It's now time to look at this model in the bigger picture. In this section, I start by assessing whether this model is effective in modelling the radio processes and reproducing observational results. I then discuss the predictions that this model makes with respect to q_{IR} . I conclude this section by discussing Shark as a flux-limited survey and how the lightcone adds to these predictions.

4.1 Is this an effective model of radio emission?

Overall, the answer to the question, "Is this a good model?" is yes, but with some caveats. Firstly, why is this a good model? The evidence presented in Figures 3.1 and 3.3 shows that this model can reproduce observational results for SFGs and (U)LIRGs. SHARK is able to reproduce the FIRC and SFR-M_{*} relation for both of these populations. This is particularly notable for (U)LIRGs which are extreme objects, which takes the model to its limits. And yet Figure 3.3 shows SHARK behaving in line with the results from Lo Faro et al. (2015).

The strongest evidence in favour of this model is shown in Figure 3.2. The left panel of this figure shows Shark able to reproduce the radio LF at z = 0 and 1.4GHz. Given the history of SAMs being unable to reproduce LFs (see Baugh et al. (2005); Lacey et al. (2016); Camps et al. (2016); Cowley et al. (2019); Trčka et al. (2020) that this model does without the need for a top heavy IMF is a reason to have confidence in it.

Further, Figure 3.11 shows that the $q_{IR} - M_*$ from Shark matches that from GAMA. It's not strictly correct to say that Shark matches the $q_{IR} - M_*$ relation from GAMA, but rather the individual galaxies from GAMA have counterparts in Shark that are within the $1 - \sigma$ od the median. Shark also agrees with results from Bell (2003) and Mao et al. (2011) (See Figure 3.12) for massive galaxies. As previously discussed, Bell (2003) did not measure the M_* of the observed galaxies, but as a flux limited survey would likely be biased towards larger galaxies $M_* \geq 10^{11} M_{\odot}$.

Shark also agrees strongly with the q_{IR} – z relationship found in Sargent et al. (2010b) for

ULIRGs, and the data from Lo Faro et al. (2015) for (U)LIRGs, with the caveta that Lo Faro et al. (2015) had a small sample size. It also finds that for massive galaxies ($10^{11} \le M_* M_{\odot} \le 10^{12}$) that LIRGs have a higher q_{IR} that ULIRGs, in line with Lo Faro et al. (2015).

Finally, Figure 3.14 shows excellent agreement with radio source counts as compiled in Tompkins et al. (2023). This an important test as all previous comparisons with observations are at a frequency of 1.4GHz. This result shows that the Bressan model reproduces the same radio source counts at different frequencies as observations as well as the Dale model. That the observations appear to peel away from the model is where the effects of AGNs start to be included. Tompkins et al. (2023) showed that the Dale model with Shark, with a corresponding AGN model is capable of reproducing the radio source counts including AGNs, so there is no reason to believe that the same model added to the Bressan model would not produce similar result. Hence the model is capable of reproducing the radio luminosities of galaxies due to star formation at different frequencies, not just at 1.4GHz to which it is calibrated.

That being said, the model does not completely reproduce all observations as strongly. The most clear is the radio LF at 150MHz at z = 0 (shown in the right panel of Figure 3.2). The Shark model over predicts the luminosity function of observations from Bonato et al. (2021b). This is a result that is out of place with other SAMs, which found that LFs usually under predict compared with observations. In my view, the dominant reason for this discrepancy is the under estimation of error in Bonato et al. (2021b). Bonato et al. (2021b) only includes Poisson error and neglects to include effects like cosmic variance or AGN contamination. If such effects were included the Shark model would likely agree with these observations.

There's a possibility that the strong agreement at z=0 for 1.4GHz, but not 150MHz is due to the spectral index used to calculate 150MHz luminosities is incorrect. The Shark model is calibrated to 1.4GHz and assumes a spectral index of $\alpha=-0.8$ to calculate other luminosities at other frequencies. A stepper spectral index (ie $\alpha=-0.9$) might produce the luminosities seen here. However such a spectral index is out of the range normally used for SFGs. Further, the model is able to produce the radio source counts at different frequencies assuming $\alpha=-0.8$, so it is unlikely that the spectral index is the primary cause of this discrepancy.

On Radio LFs, Shark also does not reproduce LFs at higher redshifts. Despite strong agreement at z=0 and 1.4GHz, Shark underpredicts the observational radio LFs from Bonato et al. (2021a); Butler et al. (2019); Novak et al. (2017); Ocran et al. (2019). This is also true for radio LFs at 150MHz for z=2,3 and 4 with only z=1 able to reproduce observational data (See Figure 3.5).

Reasons for this disagreement with radio LFs likely lies in the observations themselves. None of the observations presented calculate the true error of their measurements as they don't take into consideration cosmic variance or AGN contamination. Indeed, AGN contamination might be the primary reason for this under prediction at high redshifts. While some effort is made to remove AGNs from these observations, this is increasingly difficult at high redshifts, espeacially radio quiet AGNs. These observations are really of galaxies whose radio luminosity is dominated by star formation with some influence from AGNs, whereas the Shark model only includes the

radio emission from star formation.

Past SAMs have also encountered this problem of LFs under predicting what is seen in observations. A solution widely touted is the invoking of a top heavy IMF (Baugh et al. 2005; Lacey et al. 2008; Camps et al. 2016; Cowley et al. 2019; Trčka et al. 2020). While a top-heavy IMF would be able to reproduce the radio LFs at high z, it would disrupt the results at z = 0 and of Lagos et al. (2019). Another suggested solution from Somerville et al. (2012) is that SAMs under predict SFRs. This is particularly pertinent here since synchrotron radiation is modelled as proportional to SFR. Figure 3.6 shows that this is not the case, however as the SFR function from Shark matches that seen in observations. This leads me to conclude that the most likely explanation for this discrepancy is AGN contamination of the observational results.

Another caveat is that Shark does not reproduce the relation between qIR and M_{*} as seen in Delvecchio et al. (2021). Notably however, Delvecchio et al. (2021) does not reproduce the results from GAMA either. With GAMA having a rigorous AGN selection criteria and Delvecchio et al. (2021) using q_{IR} itself to remove AGNs, it's likely that AGNs are again the cause for this discrepancy. The result from Delvecchio et al. (2021) is like the evolution of galaxies that are dominated by star-formation, but do include some AGN emission, whereas Shark and GAMA are only SFGs. To that end, AGN contamination would also explain the evolution with redshift that other papers predict.

Overall, the model presented in this thesis shows reproduces some key observational results; the 1.4GHz radio LF, physical parameters of SFGs and (U)LIRGs, the $q_{IR} - M_*$ found in GAMA and radio source counts. However, it does not reproduce the 150MHz radio LF, the radio LF at high redshifts or the $q_{IR} - M_*$ relation found in Delvecchio et al. (2021). These discrepancies are likely due to AGN contamination within these observational results. Future work will add the AGN model to the Bressan model to investigate this possibility.

4.2 What can this model tell us about q_{IR}?

Caveats aside, this model provides important information about the FIRC for SFGs. In this section I will draw together the results of q_{IR} to discuss the overall predictions that this model has on q_{IR} .

4.2.1 Does q_{IR} evolve with M_* ?

Delvecchio et al. (2021) was the first to suggest that when evolution was seen in q_{IR} – z was actually tracing the q_{IR} – M_* instead. Shark does produce a relationship with q_{IR} – M_* , but in opposite way that is seen in Delvecchio et al. (2021). This is shown best in Figure 3.7 where galaxies above $10^{10} M_{\odot}$ are centered around $q_{IR} \sim 2.7$. However galaxies below this stellar mass are divided into two separate populations; one which is still centered around $q_{IR} \sim 2.7$,

4. DISCUSSION

but another which displays a positive correlation with q_{IR} and M_* . As best I can tell, such a population of galaxies has not been observed or predicted in any previous work.

This lower q_{IR} population has a lower L_{IR} rather than a higher L_{rad} as established in Figure 3.8. This means the dominate mechanisms behind this lower q_{IR} must produce lower L_{IR} . This is investigated in Figure 3.9, where it is established that these lower q_{IR} galaxies have (i) high Teff and are thereby dominated by birth cloud emission, (ii) less gas, (iii) are larger and more disperse, (iv) lower gas, (v) lower sSFR and (vi) lower SFR. Figure 3.10 also shows that these lower q_{IR} galaxies have less dust attenuation than higher q_{IR} .

This consequence is in line with the current understanding of $L_{\rm IR}$ emission from galaxies; young, massive stars emit UV radiation into the surrounding gas where it is re-emitted as IR radiation. However, some of these lower mass galaxies, this gas are not optically thick in the UV regime. This is because there is not enough gas to absorb and re-emit all the UV radiation being emitted. This means that these galaxies have lower $L_{\rm IR}$ and consequently a lower $q_{\rm IR}$.

 $L_{\rm rad}$ is not affected by this mechanism. At 1.4 GHz, $L_{\rm rad}$ is dominated by synchrotron radiation which arises from CR electrons accelerated by CCSNe. This process does not depend on the presence of the ISM, so $L_{\rm rad}$ is unaffected. $L_{\rm rad}$ does decrease due to decreasing amounts of SFR, but not to the same extent as $L_{\rm IR}$ does.

There are two main reasons for why this result is not supported by observations. Firstly, these galaxies are low mass and therefore low luminosity. This is compounded with the fact that these galaxies are, by definition, low L_{IR} it means that observations will struggle to observe them. Complicating this issue is the fact that in order to confirm this relationship the observation of a galaxy would need to be sufficient enough to get both L_{rad} and L_{IR} , but also enough information to find a galaxy's stellar mass. Figure 3.11 shows that even GAMA, which only looked at the local Universe, does not observe galaxies within the low q_{IR} regime because it does not observe galaxies at these low stellar masses.

Future observations by JWST and the SKA may be able to observe galaxies at such low luminosities, but current instruments are too flux limited to observe such galaxies.

The second reason is that Shark may be wrong. Previous studies predict that the linearity of q_{IR} is a conspiracy. Specifically Bell (2003) found that at low luminosities, the neither L_{IR} or L_{rad} traced star formation as well as they did at high luminosities. However, Bell (2003) found that they did so in complimentary ways, such that the linearity of FIRC is maintained.

In the Shark model, the L_{IR} is derived empirically, so is independent of SF, whereas L_{rad} is proportional to SF. Thus the result of these lower q_{IR} galaxies can be thought to confirm Bell (2003)'s conspiracy; if the Bressan model over predicts the L_{rad} at lower SFRs, then the lower q_{IR} is caused by too much L_{rad} rather than not enough L_{IR} . The problem with this argument is that this is not evident in any other result. Indeed, the biggest discrepancy that the Bressan model has is under predicting L_{rad} in the radio LF at high z.

To summarise, Shark finds no evolution in q_{IR} with M_* for galaxies that are easily observed. For lower mass and less easily observed galaxies, the median q_{IR} decreases due to a diverging population of galaxies whos L_{IR} is dominated by birth cloud emission.

4.2.2 Does q_{IR} evolve with z?

This is an area of active debate in the wider community. Some have found no evolution with z (Appleton et al. 2004; Jarvis et al. 2010; Sargent et al. 2010b,a; Bourne et al. 2011; Mao et al. 2011; Duncan et al. 2020; Thomson et al. 2014; Algera et al. 2020; Delvecchio et al. 2021) while others have found a negative correlation between q_{IR} and z (Ivison et al. 2010a,b; Magnelli et al. 2015; Delhaize et al. 2017). The predictions from Shark is that there is no evolution of q_{IR} with z. This is best shown in Figure 3.12 where, accounting for M_* , no evolution is found with z in line with results from Mao et al. (2011).

The likely explanation of this is AGN contamination at high redshifts. At high redshifts, the removal of AGNs is difficult due to the lack of clarity in the data and inability to measure all properties. As such AGNs, especially radio quiet AGNs, are likely still contained within the data used to measure q_{IR} , thus dragging q_{IR} down.

In Future work, I will investigate the extent to which AGN contamination affects the q_{IR} relation by including AGN radio emission in this model.

Chapter 5

Conclusion

In this dissertation I have introduced a model of radio emission from star formation to the semi-analytic model of galaxy formation Shark. I uses the approach developed in Bressan et al. (2002) and model synchrotron and free-free emission separately. Synchrotron is modelled as proportional to the rate of CCSNe with a minor contribution from SNe remnants. Free-free emission is modelled as proportional to the production rate of Lyman alpha photons. I also utilise the results from Lagos et al. (2019) which successfully modelled the UV-FIR emission using Shark to find the IR emission of galaxies.

This model is capable of reproducing (i) the 1.4GHz radio LF at z = 0, (ii) physical parameters of SFGs and (U)LIRGs including the FIRC and SFR – M_* relationship, (iii) the q_{IR} – z relationship for (U)LIRGs (iv) the q_{IR} – M_* when compared with data from the GAMA survey and Bell (2003) (v) the q_{IR} – z compared with Mao et al. (2011) and (iv) radio source counts. However it does not reproduce the 1.4GHz LF at high redshifts or the 150MHz radio LFs. This is likely due to AGN contamination within the observational data. It is unlikely that a top-heavy IMF is required or that the SFR in Shark is lower than observations.

Using this model I test the q_{IR} relationship with M_* and z. Overall, I find that q_{IR} evolves primarily with M_* and not z. This is only true for galaxies with $M_* < 10^{10} M_{\odot}$, above which no evolution is seen. Below this mass limited I find two divergent populations of galaxies, one with no evolving q_{IR} and one with lower q_{IR} . The lower q_{IR} population of galaxies is dominated by birth cloud emission and is not optically thick to UV emission leading to lower L_{IR} . Such galaxies have not been observed but there's a possibility they might with the SKA and JWST. However, the breakdown of the linearity of FIRC may confirm the result of Bell (2003), that low L_{rad} do not trace SFR as well as higher L_{rad} .

A lightcone is also made using this model. This is used to confirm the radio source counts form Tompkins et al. (2023). It's also used to show the properties of observable high redshift galaxies in flux limited surveys. At high redshifts fewer galaxies are observed and those that are have a higher mass and higher SFR. The need for K-correction is displayed in the stratification of these results; increasing frequencies see fewer galaxies. This offers an application of this model as a way to test the bias of particular flux limited surveys towards particular galaxy properties.

Overall this model provides an opportunity to investigate the assumptions that are made with respect to star formation in galaxies. The success of this model means that the proportionality between $L_{\rm rad}$ and SFR is a safe assumption. It also offers itself as a tool for future galaxy surveys to predict the radio continuum based on the physics assumed here.

Chapter 6

Future Work

There are three possible extensions to the work that I have presented in this thesis; running the model on Shark V2.0, extending the model to include AGN emission and extending the model to include further emission processes at different frequencies.

This thesis was completed on Shark V1.0 which is outlined in Lagos et al. (2018). However, the next version of Shark, Shark V2.0 has been in development for sometime. The major difference between the two versions is an improved method of calculating SFR (Lagos et al. in prep). This has an obvious impact on this model of radio emission since SFR is the main driver of synchrotron radiation here. Further, there are a number of galaxies in Shark V1.0 where the SFR is not converged and thus excluded from this and other investigations. If Shark V2.0 improves these converged SFRs, then these galaxies could be included in the sample.

As mentioned throughout this thesis, this model of radio emission only includes emission from star formation. It does not include any contribution from AGNs. This has lead to complications in comparing with observations that have not or cannot remove AGNs. The model presented here is of an incomplete radio emission. In order to complete it a model of AGNs must be added. To do that I would add the Griffin et al. (2019) model of AGN emission. This model uses the mass of the black hole and accretion rate of each galaxy to model the total radio emission from AGNs. These parameters are readily available in Shark.

Tompkins et al. (2023) successfully used this model with the Dale model to use Shark to reproduce the radio source counts. There is little reason to expect that combining this model with the Bressan model will yield different results. Introducing AGNs to the radio emission model would complete the radio SED and allow for investigations into the entire radio spectrum. This includes the extent to which AGN contamination occurs within observations.

Finally, the model presented in Shark could be extended to higher frequencies like X-Rays. Lagos et al. (2019) presented the Shark model of UV to the IR. This thesis has introduced the radio emission from star formation. What remains of the electromagnetic spectrum includes X-Rays and Gamma Rays. There exists a few models of X-Ray emission that are capable of being used in SAMs.

Recently, Zhong et al. (2023) proposed a model of X-Ray emission based on the local gas density

of ionised gas based on the results of the radial density profiles. This method is a little involved for Shark and requires a working model of AGN feedback. This is presented as an improved model from a simpler model which uses the cooling rate within a cooling radius of the galaxy to model X-Ray emission. A galaxy's cooling rate is directly tracked in Shark, so this simple model could be easily implemented.

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