

The growth of supermassive black holes in galaxies: co-evolution and regulation of the star formation activity

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Abstract

In this Thesis we study how the accretion of the central supermassive black hole evolves in galaxies in all their star formation life phases and through a wide range of cosmic epochs.

We take two complementary approaches. First, we perform a statistical study on a large sample of galaxies from the COSMOS field where we take advantage of X-ray Chandra data to estimate black hole accretions, via a combination of stacked data and individual detections, and compare them with their star formation properties, estimated from far-infrared emission combined with ultra-violet emission. Then, we use semi-empirical models to create galaxy mock catalogs onto which we perform an analogous analysis in order to pin down which parameters control the black holes' X-ray emission and its evolution.

We find a picture in which the bulk of the black hole and stellar masses are accreted in the star forming phase through secular processes, where the average black hole accretion follows a relation with stellar mass similar to the "main sequence", i.e. the relation between the star formation rate and the stellar mass followed by star forming galaxies, having a similar evolution in time but with a more efficient accretion at high stellar masses. The starburst phase appears to have a significant enhancement of the SFR but a lesser impact on the black hole accretion, which has a smaller enhancement especially at high redshift. Quiescent galaxies, on the other hand, undergo a significant decline in their star formation, while the black hole accretion is still noticeable. This observed evolution of the X-ray luminosity with time and galaxy phase is compatible with a change in the average Eddington ratio but is mostly independent on the duty cycle. We find a super-linear relation between black hole and stellar mass which, in order to reproduce the observations, should be combined with an average Eddington ratio that depends on stellar mass. Our results point in the direction of galaxy downsizing, i.e. a fast accretion of the black hole and stellar mass at very high redshift for the most massive galaxies, followed by a steep decrease in accretion, while low-mass galaxies accrete their mass more slowly, with an accretion rate that

decreases more slowly with time.

On a separate note, we study the gas distribution in a sample of local active galaxies, by analyzing their continuum and reflected X-ray light curves and reproducing the observed damping of the variations of the reflected component through Monte Carlo simulations.

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CHAPTER 1

Introduction

Our Universe is well described by the λ CDM model. In this model the universe is composed by baryonic matter, dark matter and dark energy. We can only observe the baryonic matter in the Universe, which makes up only $\sim 5\%$ of its content, while dark matter is $\sim 27\%$ and dark energy $\sim 68\%$. In this model the chemical composition of the Universe determined by the cosmological nucleosynthesis which takes place in the first minutes of its life, and is mostly of hydrogen and helium (gas) and very few heavier elements (*metals*). As the universe expands in time, dark matter falls into the potential wells arisen from the cosmic fluctuations, thereby dragging baryonic matter along with it. Dark matter will collapse to form halos, while baryonic matter will collapse to form stars and galaxies (Komatsu et al., 2009; Spergel et al., 2007; Planck Collaboration et al., 2020). It is commonly believed that in the center of each galaxy lies a supermassive black hole (SMBH, or simply BH), an object whose gravity potential is so strong that not even light can escape it. This belief has been recently consolidated by studying the motion of the stars orbiting the center of our own galaxy, the Milky Way, which confirmed that a mass as high as $\sim 4.3 \times 10^6 M_\odot$ is confined within the central 0.1 arcsec (Ghez et al., 2008; Gravity Collaboration et al., 2019, 2021), and even more so by the direct observations of the Event Horizon Telescope of the center of M87, a local galaxy with a very active SMBH (Event Horizon Telescope Collaboration et al., 2019).

While it is easy to describe the behavior of dark matter, as it only interacts gravita-

tionally, the same cannot be said of baryonic physics, which rules galaxy evolution. As the stars evolve they produce more and more metals, giving rise to dust and creating older stellar populations. Stars interact with their environment through strong stellar winds and can explode as supernovae. More gas supplies can flow into the galaxy from the intergalactic medium, the gas can cool down to form new stars, but also fall into the central potential well of the galaxy to feed the SMBH. The galaxy itself can undergo morphological evolution as it interacts with its environment and/or through mergers with other galaxies.

In this thesis we are going to study the evolution of the gas accretion by tracing the growth of BH, i.e. the black hole accretion rate (BHAR), and the accretion of new stellar populations, measured as the star formation rate (SFR) as a function of time and of the life phase of the galaxy, defined by how actively it is forming new stars.

The rapid accretion of the SMBH in the core of the galactic center, that defines an active galactic nucleus (AGN), is usually observable at various wavelengths, since it involves, directly or indirectly, a large number of physical mechanisms. Nowadays, the differences among AGN showing different observational characteristics are explained in the context of the unified model (Rowan-Robinson, 1977; Antonucci, 1984; Antonucci & Miller, 1985; Antonucci, 1993; Urry & Padovani, 1995; Netzer, 2015), as due to the orientation with respect to the observer.

As already mentioned, galaxies interact with their medium. They can undergo violent interactions or mergers that can end up causing a morphological change or they can evolve in a more passive sort of way in which they consume the gas that is already present *in situ* and may receive a slow gas inflow from the cosmic web of filaments containing pristine gas. The latter case is what we call secular evolution and in this case, the SFR of a galaxy seems to be regulated by a simple empirical relation: the bigger the galaxy is, in terms of stellar mass M_* , the higher the SFR is. This law, usually referred to as main sequence (MS) of star-forming galaxies (Brinchmann et al., 2004; Elbaz et al., 2007; Daddi et al., 2007; Noeske et al., 2007; Rodighiero et al., 2014; Aird et al., 2017), seems to be valid for local as well as for distant galaxies, for a wide range of stellar masses, and considering different SFR tracers. A galaxy can be considered as “active” by the point of view of the star formation, when its SFR is consistent or higher than the main sequence. A galaxy is then “passive” (or equivalently “quiescent”) when the SFR is very low or absent. At the opposite side of the main sequence, a special class of star forming galaxies is represented by the so called “starbursts”, which show SFRs even ten times, or higher than that of the main sequence. These rare

objects, represent a very peculiar and still not well known phase of galaxy evolution.

1.1 SFR and its tracers

Stars convert their mass into light and energy through nuclear fusion. From the integrated light emitted by the stars in a galaxy, there are many ways in which the SFR can be inferred. Calibrations of SFR indicators have been presented in the literature for almost 30 years, derived across the full electromagnetic spectrum, from the X-ray, through the ultraviolet (UV), via the optical and infrared (IR), all the way to the radio, and using both continuum and line emission. Extensive reviews of this topic are reported in, e.g., Kennicutt (1998); Kennicutt & Evans (2012); Vattakunnel et al. (2012). The basic goal is to identify emission that probes newly or recently formed stars, while avoiding as much as possible contributions from evolved stellar populations.

In unresolved systems, SFR indicators are merely measures of luminosity, either monochromatic or integrated over some wavelength range, with the goal of targeting continuum or line emission that is sensitive to the short-lived massive stars. The conversion from the luminosity of massive stars to a SFR is performed under the assumption that: (1) the star formation has been roughly constant over the time-scale probed by the specific emission being used; (2) the stellar initial mass function (IMF) is known (or is a controllable parameter) so that the number of massive stars can be extrapolated to the total number of high+low mass stars formed; and (3) the stellar IMF is fully sampled, meaning that at least one star is formed in the highest-mass bin, and all other mass bins are populated accordingly with one or more stars.

SFR indicators in the UV/optical/near-IR range ($\sim 0.1\text{--}5 \mu\text{m}$) probe the direct stellar light emerging from galaxies, while SFR indicators in the mid/far-IR ($\sim 5\text{--}1000 \mu\text{m}$) probe the stellar light reprocessed by dust. In addition to direct or indirect stellar emission, the ionizing photon rate, as traced by the gas ionized by massive stars, can be used to define SFR indicators; photo-ionized gas usually dominates over shock-ionized gas in galaxies or large structures within galaxies (e.g., Calzetti et al., 2004; Hong et al., 2011). Tracers include hydrogen recombination lines, from the optical, through the near-IR, all the way to radio wavelengths, forbidden metal lines, and, in the millimeter range, the free-free (Bremsstrahlung) emission. The X-ray emission produced by high-mass X-ray binaries, massive stars, and supernovae can also be used to trace SFRs (Aird et al., 2017). Finally, the synchrotron emission from galaxies can be calibrated as a SFR indicator (Condon, 1992), since cosmic rays are produced and

accelerated in supernova remnants, and core-collapse supernovae represent 70% or more of the total supernovae in star-forming galaxies (Boissier & Prantzos, 2009).

The SFR efficiency measures the fraction of cold gas, i.e. gas in the molecular phase, which is then transformed into stars in a free fall time, which is the time scale of the collapse of a molecular cloud. Typical values of this efficiency are $\epsilon_{SF} \approx 10^{-3}$.

1.2 AGN and their emission mechanisms

Active galactic nuclei are the ensemble of physical and observational phenomena that occur at the very center of galaxies and ascribable to the presence of a SMBH accreting matter at high rates. As can be seen in Fig. 1.1, black solid line, their emission encompasses the whole electromagnetic spectrum and cannot be explained by stars, gas and dust alone but requires high temperature and highly energetic physical processes where about 10% of the mass accreted is converted into radiation. In fact, when compared to the spectrum of a starburst galaxy (gray solid line in Fig. 1.1), one can see that even if this kind of galaxy hosts the most powerful stellar emissions, the ongoing physical processes in their interior cannot reach the extremes of the electromagnetic spectrum like AGN do, and have a different spectral shape. Luminosity variability measurements allow to determine the size of the AGN sources that are not bigger than few parsecs. AGN are the most powerful emitting sources in the universe and for quite a long time, they were the only kind of source that was detectable at high redshifts. Nowadays, they represent one of the important mechanisms, together with the SF, that seem to drive the evolution of galaxies and the environment in which they evolve.

The infalling gas will produce an optically thick disk of material, an accretion disk on a subparsec scale, that will emit thermally because of its own viscosity (e.g. Shakura & Sunyaev, 1973; Rees, 1984). The temperature of the gas will increase as it approaches the SMBH and in a typical AGN will range at $T \approx 10^3 - 10^4$ K corresponding to a majority of the emission at $\approx 30 - 300$ nm (blue line in Fig. 1.1). Around the accretion disk there is a geometrically and optically thick warm-hot dusty and molecular torus which is at a luminosity dependent distance, but within the gravitational influence of the SMBH. It may be considered as the extension of the accretion disk at a distance where dust and molecules can form. The torus is heated by the absorption of shorter-wavelength photons from the accretion disk which it will in turn then re-emit thermally at lower energies, at NIR and MIR wavelengths (red dashed line in

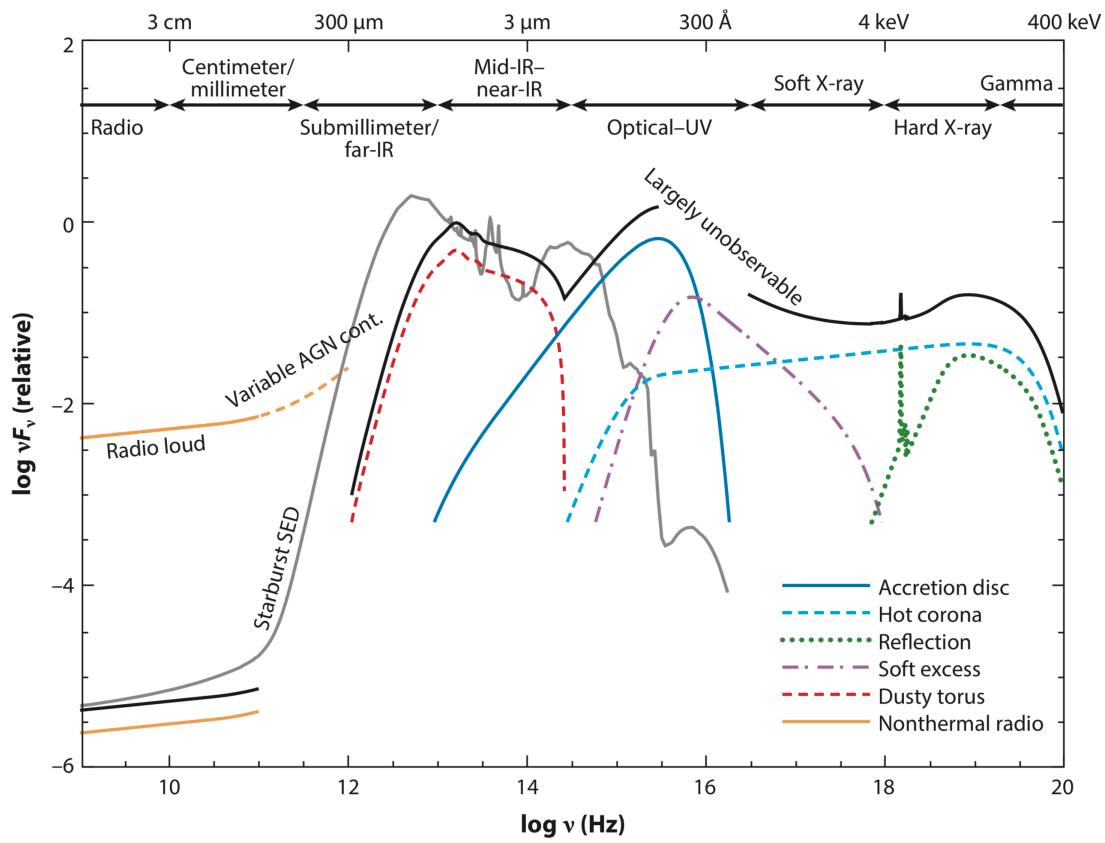


Figure 1.1: Spectral energy distribution of an AGN. Figure from Hickox & Alexander (2018)

Fig. 1.1). Another important source of emission from the black hole is a hot corona above the central part of the accretion disk. This emits in the X-rays through inverse Compton scattering of photons from the accretion disk with relativistic electrons and has a power law shape $N(\nu) \propto \nu^{-\Gamma}$, with $\Gamma = 1.8 \pm 0.2$ (light blue dashed line in Fig. 1.1). The relativistic electrons in the corona and also in large-scale radio jets can generate synchrotron radiation observable in the radio (yellow line in Fig. 1.1). The primary X-ray continuum can be reprocessed via Compton scattering and photoelectric absorption, thus leading to two important features in the X-ray spectrum: the Fe K α emission line and the so-called “Compton-hump”, together known as the AGN “reflection” component (green dotted line in Fig. 1.1).

In the optical bands, the observed spectrum strongly depends on the inclination of the object. If the dusty torus inclination allows to observe the inner regions we will see emission lines from the broad line region, which is a high density region at about 0.01-0.1pc from the SMBH, close to the accretion disk, where the gas reaches speeds of thousands of km/s. This inclination question makes the torus the key ingredient to explain the main dichotomy between obscured and unobscured AGN. The narrow line region is more external, within the central kiloparsec (e.g. Humphrey et al., 2015; Villar-Martín et al., 2016), and is less dense, therefore allowing for forbidden line transitions with lower velocity dispersions, of a few 100s of km/s.

1.3 The connection between SFR and BH accretion

Observational studies and cosmological simulations have revealed a deep interconnection between galaxies and their central supermassive black hole (SMBH): they appear to coevolve and affect each other during their lives. It has been reported several times that the SMBH mass (M_{BH}) correlates with a number of galaxy properties, including bulge mass, stellar velocity dispersion, and starlight concentration as quantified for example by the Sérsic index. The existence of these relations is not trivial to explain since these relations hold for several orders of magnitude in both SMBH mass and the host parameter and the SMBH has a very small sphere of influence compared to the galaxy size. While it is relatively easy to explain that gas inflows from the intergalactic medium can feed the star formation during secular processes, it is not easy to understand how the gas loses angular momentum and funnels to the center of the galaxy to feed the SMBH (Storchi-Bergmann & Schnorr-Müller, 2019). Bernardi et al. (2007) have seen that a selection effect caused samples of AGN galaxies to be bi-

ased to higher velocity dispersions σ at a given optical or NIR luminosity L . By modeling this bias they were able to show that the most fundamental relation is between $M_{\text{BH}} - \sigma$ with a very low intrinsic dispersion, while the relation between $M_{\text{BH}} - L$ is a consequence of a relation between $\sigma - L$ and is therefore more biased, as it overpredicts abundances of massive BHs. Also Shankar et al. (2017) showed that $M_{\text{BH}} - \sigma$ is the most fundamental in the scaling relations between black holes and galaxies by studying the residuals of the relations between M_{BH} and σ , Sérsic index n and M_{bulge} in SDSS galaxies. Finally, Monte Carlo simulations have shown that a bias can be introduced when dynamically measuring M_{BH} : in order to measure black hole masses, the black hole sphere of influence must be resolved, but because of instrumental limitations it is not possible yet to probe the sphere of influence of the smaller (and also farther) SMBHs, thus resulting in an increased $M_{\text{BH}} - \sigma$ relation of at least a factor three (Shankar et al., 2016).

Interesting insight in this respect has been gained by studying the cosmic star formation rate density (SFRD) and black hole accretion rate densities (BHARD), which are the amount of star formation and black hole accretion per unit of comoving volume at a given redshift¹. With the fundamental contribution from deep and wide field surveys it has been possible to study them thoroughly and to unveil their evolution. As can be seen in Fig. 1.2, the BHARD has been determined from IR observations with *Herschel* (Pilbratt et al., 2010) and from X-ray observations with XMM-Newton (Jansen et al., 2001) and Chandra (Weisskopf et al., 2000). In the IR it was possible to determine the BHARD up to $z \sim 3$ (Delvecchio et al., 2014), while in the X-rays, thanks to wide and deep surveys it was possible to determine it up to $z \sim 6$ (Vito et al., 2018). The history of star formation in galaxies throughout the life of the Universe has been thoroughly constrained in recent years: through infrared data with *Herschel*, ultraviolet data with *Galaxy Evolution Explorer* (GALEX) (see Madau & Dickinson, 2014, for a review) and more recently complemented by sub-mm ALMA data, especially at high redshift up to $z \sim 10$ (Bouwens et al., 2020; Gruppioni et al., 2020). All of these studies have shown that both BHARD and SFRD reach a peak of activity at redshift $z \sim 2$ and then decrease to the present epoch (Boyle & Terlevich, 1998).

Additionally, semianalytic models and hydrodynamic simulations show that a self-regulating mechanism, the so called feedback, is required between the star formation and the black hole accretion in order to reproduce local scaling relations, i.e.

¹The comoving volume is a measure of volume that factors out the expansion of the universe, resulting in a volume that does not change in time due to the expansion of space.

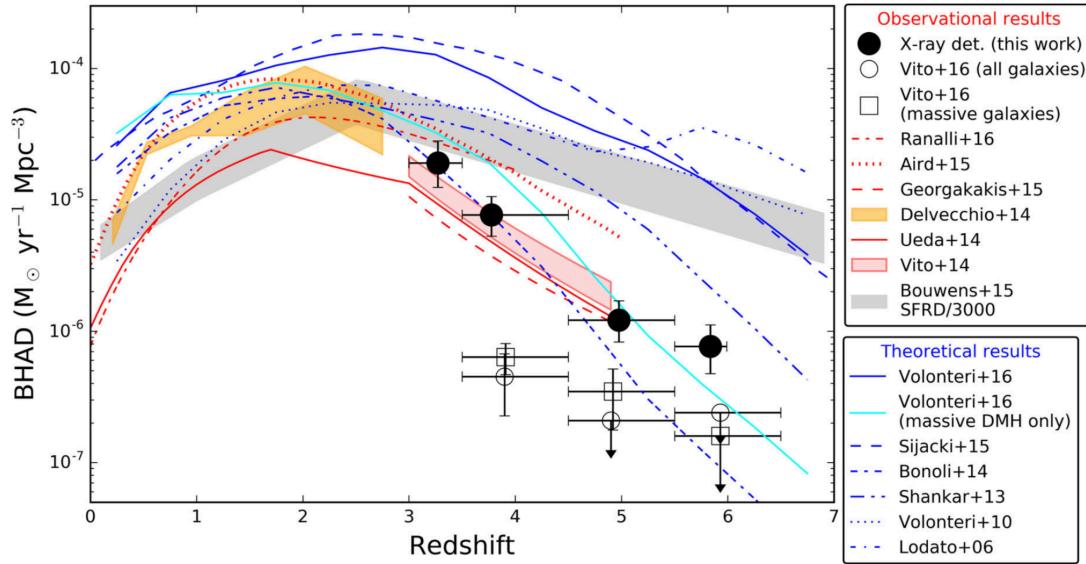


Figure 1.2: BHARD from Vito et al. (2018) and other works as reported in the legend. SFRD from Bouwens et al. (2015) is also reported for comparison purposes.

relations that describe strong trends that are observed between important physical properties of galaxies (see Somerville & Davé, 2015, for a review). Since the pioneering models of galaxy evolution in a cold dark matter framework it was clear that there was an “overcooling problem”: most of the gas should have cooled and condensed into stars by the present day while it is observed that less than 10% of it is in the shape of stars, so some suppression of cooling and star formation had to be introduced. This could be as energy released by supernovae (Larson, 1974; White & Rees, 1978; Dekel & Silk, 1986; White & Frenk, 1991) and the impact of this phenomenon would be important in regulating star formation in low mass galaxies. On the other hand, luminosity functions predict more high luminosity galaxies in the local universe than we observe, and this might be prevented from happening through feedback from the SMBH as it actively accretes gas .The accretion of gas in the BH is an very efficient process, and a fraction of $\approx 5 - 45\%$, depending on the spin of the black hole (e.g. Kerr, 1963; Shapiro & Teukolsky, 1983), will be emitted as electromagnetic radiation, thereby allowing for very high AGN luminosities. Simple calculations have shown that when the black hole becomes sufficiently massive ($> 10^7 M_\odot$) its maximum accretion luminosity, the Eddington luminosity, becomes high enough that its winds could in principle blow out all of the gas in the entire galaxy (Silk & Rees, 1998). This feedback mode, called quasar mode, prevents star formation by delivering momentum to the galactic gas

and removing it. There is also a radio mode in which the BH is in a low accretion mode and heats the gas, thereby preventing it from collapsing and forming new stars (Harrison, 2017).

The decline of the SFRD at low redshift is thought to depend on the decreasing availability of cold gas that is required to form stars and accrete onto black holes (e.g., Feldmann et al., 2016). Since most of the galaxies lie on the main-sequence of star-forming galaxies, they are the ones that dominate the SFRD evolution (Rodighiero et al., 2011). The observation of the main sequence at all redshifts and its lack of evolution of the slope, but only an increase of the normalization with redshift, reinforces the belief that most of the SF happens through secular processes (Tasca et al., 2015; Schreiber et al., 2015; Tomczak et al., 2016): at earlier epochs, galaxies of a given stellar mass were forming more stars than in the local Universe. Starburst galaxies are instead a small fraction of star-forming galaxies that are undergoing a stochastic and short-lived episode of considerable galaxy growth, and their SFRs and gas fractions are higher than those on the MS at the same stellar mass, but they cannot be dominating the SFRD because of their low numbers ($\sim 2\%$ Rodighiero et al., 2011). Quiescent galaxies, on the other hand, have little star formation but have been shown to evolve from $z \sim 1.8$, where they have significant amounts of dust and gas ($\sim 5 - 10\%$) but their star formation efficiency is low, to the local universe, where they are gas poor (Gobat et al., 2018).

Just like star formation, black hole accretion depends on the availability of cold gas. In a classic evolutionary picture predicted by simulations one could have a star forming disk galaxy on the main sequence which undergoes a major merging event (Di Matteo et al., 2005; Hopkins et al., 2008): During the early stages of galaxy merging, gas can efficiently cool and lose angular momentum, eventually feeding black hole growth and central star formation, producing a starburst. The black hole accretion is initially obscured by thick layers of dust, which are then possibly removed by its increasing radiation and momentum feedback, revealing the quasar. Eventually, the gas is consumed, the quasar luminosity fades rapidly, and the star formation episode ceases. This leaves a “red and dead” elliptical galaxy with no or very little star formation or black hole accretion (Granato et al., 2004; Lapi et al., 2006). This picture seems to be validated by low redshift (ultra) luminous IR galaxies ((U)LIRGs) which are starburst galaxies often undergoing a major merger event (Sanders & Mirabel, 1996), but it doesn’t seem to apply to higher redshift starbursts, as many of these don’t show signs of a disturbed morphology (Elbaz et al., 2007; Rodighiero et al., 2019).

Other theoretical models have predicted pictures of an in situ co-evolution scenario in which SF and black hole accretion are fed by the gas already present in the galaxy and triggered by the early collapse of the host dark matter halos (e.g. Lapi et al., 2006, 2011, 2014; Lilly et al., 2013; Aversa et al., 2015a; Mancuso et al., 2016) and/or steady cold gas streams along filaments of the cosmic web (e.g. Dekel et al., 2009; Bournaud et al., 2011). The accretions are subsequently controlled by self-regulated baryonic physics and in particular by energy feedback from supernovae (SNe) and AGN. In this picture starbursts are gas rich young galaxies, placed to the left of the main sequence (as opposed to *above*). These galaxies then evolve at about a constant SFR while increasing their stellar masses, thus reaching the main sequence locus (Mancuso et al., 2016; Lapi et al., 2018). Additionally Wang et al. (2019) find in their semi-analytic models of galaxy formation that not all galaxies in the starburst locus have recently undergone a major merger and that not all the galaxies which have undergone a major merger event result in a starburst. Observational results are still not capable to discern between these evolutionary scenarios as deep high resolution data for large statistical samples of starbursts are needed, and only future facilities will be capable to provide them.

These pictures might suggest that just like the majority of galaxies follow a main-sequence in the SFR- M_* plane, a similar relation between the black hole accretion rate (BHAR) and the M_* might exist: the more massive the galaxy, the higher the availability of inflowing gas for star formation and black hole accretion, which would mean that they both should correlate with stellar mass. In addition, galaxies offset from the main-sequence (starbursts and quiescents), might have a BHAR that varies accordingly with the gas that is typically available in that phase (Rodighiero et al., 2019).

In order to search for these potential correlations, many authors have used the X-ray luminosity of galaxies as a proxy of BHAR: X-rays are very energetic photons that are created very close to the central SMBH, and other contaminants in the host galaxies at these wavelengths, for example, emission from stellar processes or binary systems, are usually less powerful and not dominant (e.g. Brandt & Alexander, 2015). Nevertheless, the first studies that traced the instantaneous BHAR with X-ray flux failed at finding any BHAR- M_* relation (Silverman et al., 2009; Shao et al., 2010; Mullaney et al., 2012b; Rosario et al., 2012; Azadi et al., 2015). A lack of a correlation between SFR and BHAR does not by itself necessarily imply a lack of physical connection. It might arise, for example, from different duty cycles and variabilities that characterize the

two processes. Episodes of star formation last for several Gyr covering a scale of Kpc, while the SMBH duty cycles are believed to be very short, with accretion episodes of about 10^5 yr and variability timescales that range from minutes to months.

In order to constrain more robust and reliable BHAR, it is thus necessary to average their growth rate over a long time interval. A very promising technique to achieve this goal consists of stacking X-ray images. Since the BHAR is a stochastic event, stacking large samples of galaxies in a given volume, by grouping them based on optical properties, e.g. the stellar mass, is equivalent to averaging the growth rate of all galaxies. Furthermore stacking allows us to perform studies on mass-complete samples by averaging the count rates of the X-ray images in the optical positions of the galaxies, thus increasing the signal-to-noise ratio, and allowing us to reach fluxes well below the single-source detection threshold of the observations. Previous works indeed searched for a relation between M_* and average BHAR by stacking X-ray images (e.g., Mullaney et al., 2012a; Rodighiero et al., 2015; Yang et al., 2017), and the probability distribution of specific X-ray luminosity (X-ray luminosity divided by galaxy stellar mass) with a maximum likelihood approach (Aird et al., 2012; Bongiorno et al., 2012; Yang et al., 2018) and a Bayesian approach (Aird et al., 2018). All studies point toward a positive correlation between the BHAR and the M_* for star-forming galaxies, very similar to the main-sequence of star-forming galaxies, with a slope close to unity, non-negligible redshift evolution, a positive slope for the BHAR-to-SFR ratio as a function of stellar mass, and indications of different behaviors for quiescent and starburst galaxies. Thus far, no study has presented a complete analysis throughout all galaxy life phases by highlighting the evolution of the accretions and the BHAR-to-SFR ratio throughout cosmic time. This is what we present here, in order to understand if the efficiencies of gas conversion in the black hole correspond to those of the star formation, how they vary with time, stellar mass and especially with galaxy life phase. So far, only Aird et al. (2019) have shown that the fraction of AGN galaxies is higher below the main-sequence and in starbursts than in the main sequence.

We here characterize the evolution of the average BHAR for normal star-forming, quiescent, and starburst galaxies at $0.1 < z < 3.5$. This redshift interval encompasses the majority of the history of the Universe and contains two crucial epochs in its evolution: the peak of the star formation rate density and BHAR of the Universe at $z \sim 2$, and their decline to the local Universe. In Chapter 2 we take advantage of the unique depth, area, and wavelength coverage of the COSMOS field, which allows us to select a mass-complete sample with large statistics out to very high redshifts for each galaxy

phase. This is particularly important for starbursts, which are rare objects and require a large field in order to be found in good numbers for statistics. Then in Chapter 3 we use the unique flexibility and simplicity of semi-empirical models to generate mock galaxy catalogs and perform a similar analysis in order to explore the parameters that rule the $\langle L_x \rangle - M_*$ by also gaining a more comprehensive view of how BHs accrete at different epochs and in different host galaxies.

1.4 Gas in active galaxies in the local universe

On a different approach, in Chapter 4 we study the distribution of the gas generating the X-ray reflected component (green dotted line in Fig. 1.1) in a sample of local active galaxies, by assuming that the reflected emission comes from the X-ray continuum and is reprocessed by a dense layer of gas which will produce Fe K α emission and damp the variations of the continuum based on its size. We work with a sample of galaxies from Andonie et al. (in prep.) for which we have light curves of both the continuum and Fe K α line extracted from Chandra and XMM-Newton archival data. We run Monte Carlo simulations of these light curves, modeled as red noise, in order to determine the size of the reprocessing gas based on the observed smoothing of the Fe K α light curves.

CHAPTER 2

An observational point of view

The work in this chapter has been published on Astronomy and Astrophysics.

In this chapter, we study the coevolution between the black hole accretion rate (BHAR) and the star formation rate (SFR) in different phases of galaxy life: main-sequence star-forming galaxies, quiescent galaxies, and starburst galaxies at different cosmic epochs. We exploited the unique combination of depth and area in the COSMOS field and took advantage of the X-ray data from the *Chandra* COSMOS-Legacy survey and the extensive multiwavelength ancillary data presented in the COSMOS2015 catalog, including in particular the UVista Ultra-deep observations. These large datasets allowed us to select a mass-complete sample onto which we perform an X-ray stacking analysis and combine it with detected sources in a broad redshift interval ($0.1 < z < 3.5$) with unprecedented statistics for normal star-forming, quiescent, and starburst galaxies. The X-ray luminosity was used to predict the BHAR, and a similar stacking analysis on far-infrared *Herschel* maps was used to measure the corresponding obscured SFR for statistical samples of sources in different redshifts and stellar mass bins.

We focus on the evolution of the average SFR-stellar mass (M_*) relation and compare it with the BHAR- M_* relation. This extends previous works that pointed toward the existence of almost linear correlations in both cases. We find that the ratio between

BHAR and SFR does not evolve with redshift, although it depends on stellar mass. For the star-forming populations, this dependence on M_* has a logarithmic slope of ~ 0.6 and for the starburst sample, the slope is ~ 0.4 . These slopes are both at odds with quiescent sources, where the dependence remains constant ($\log(\text{BHAR}/\text{SFR}) \sim -3.4$). By studying the specific BHAR and specific SFR, we find signs of downsizing for M_* and black hole mass (M_{BH}) in galaxies in all evolutionary phases. The increase in black hole mass-doubling timescale was particularly fast for quiescents, whose supermassive black holes grew at very early times, while accretion in star-forming and starburst galaxies continued until more recent times. Our results support the idea that the same physical processes feed and sustain star formation and black hole accretion in star-forming galaxies while the starburst phase plays a lesser role in driving the growth of the supermassive black holes, especially at high redshift. Our integrated estimates of the M_* - M_{BH} relation at all redshifts are consistent with independent determinations of the local M_* - M_{BH} relation for samples of active galactic nuclei. This adds key evidence that the evolution in the BHAR/SFR is weak and its normalization is relatively lower than that of local dynamical M_* - M_{BH} relations.

2.1 Data

Our study focuses on the relation between average BHAR, SFR, and M_* across a wide redshift range, from $z = 0.1$ to $z = 3.5$, and across different evolutionary stages of galaxies: (i) normal star-forming galaxies, which accrete gas secularly to slowly form new stars, and in which the bulk of the cosmic star formation took place; (ii) starburst galaxies, which are a small fraction of all galaxies at all epochs (Rodighiero et al. 2011, $\sim 2\%$; but see also Caputi et al. 2017) and are going through a great burst of star formation that drives them significantly above the main-sequence galaxies on the SFR- M_* plane; and (iii) quiescent galaxies, which form stars at a very slow pace.

Star-forming and quiescent galaxies can be easily selected at all redshifts from their emission in the optical/near-infrared (NIR) rest-frame bands, whereas starburst galaxies are heavily obscured in the optical and are therefore more easily identified by the thermal emission from the dust in the far-infrared (FIR). Therefore we selected our samples of star-forming and quiescent galaxies from the catalog by Laigle et al. (2016), which is NIR selected, while we used the FIR-selected catalog by Gruppioni et al. (2013) to identify starburst galaxies. For the BHAR we used the catalog by Civano et al. (2016), obtained from the *Chandra* COSMOS-Legacy program, and complemented it by

stacking on their X-ray images. We estimated the SFR for star-forming and quiescent galaxies by combining FIR stacking on *Herschel* images and detections with UV luminosity. We present in this section the catalogs we used for the sample selection and the data we used to estimate the average BHAR and SFR.

2.1.1 Optical-NIR catalog

We selected our sample of star-forming and quiescent galaxies in the COSMOS field (Scoville et al., 2007) from the COSMOS 2015 catalog (Laigle et al., 2016), which uses photometry from UV (*Galex*) to the mid-infrared (MIR, IRAC). This catalog constitutes the UltraVISTA DR2.

The multiwavelength catalog includes UV photometry in the far-UV and near-UV (NUV) bands with the *GALEX* satellite (Zamojski et al., 2007); UV/optical photometry in the u^* -band from the Canada-France-Hawaii Telescope (CFHT/MegaCam); optical photometry from the COSMOS-20 survey, which is composed of 6 broad bands (B, V, g, r, i, z+), 12 medium bands (IA427, IA464, IA484, IA505, IA527, IA574, IA624, IA679, IA709, IA738, IA767, and IA827), and 2 narrow bands (NB711 and NB816), taken with the Subaru Suprime- Cam (Taniguchi et al., 2007, 2015); new and deeper z^{++} and Y-band data, both taken with the Hyper-Suprime-Cam (HSC) on Subaru; H and K_s NIR photometry obtained with WIRcam/CFHT (McCracken et al., 2010), as well as deeper J, H, and K_s imaging obtained with VIRCAM/VISTA UltraVISTA (McCracken et al., 2012) in the central 1.5 deg² of the COSMOS field (the coverage by UltraVISTA is not homogeneous, with alternating "deep" and "ultradeep" stripes that reach depths of 24 and 24.7 in K_s -band, respectively); NIR data from IRAC, as part of the SPLASH COSMOS , together with S-COSMOS (Sanders et al., 2007); MIR and FIR data with *Spitzer* IRAC and MIPS (Multi-band Imaging Photometer), from the *Spitzer* Extended Mission Deep Survey and the *Spitzer*-Candels survey (Ashby et al., 2015) data, among others; FIR from the *Herschel* PACS and SPIRE instruments, taken as part of the PACS Evolutionary Probe (PEP) guaranteed-time key program (Lutz et al., 2011), the largest field of the program, observed for about 200 h to a 3σ depth at 160 μm of 10.2 mJy and at 100 μm \sim 5 mJy.

The point spread function (PSF) of the VISTA bands was estimated in each photometric band by modeling isolated known stars from the COSMOS ACS/HST catalog (Koekemoer et al., 2007; Leauthaud et al., 2007) with the PSFEX tool (Bertin, 2013); The target PSF was chosen in order to minimize the applied convolutions, and it is the

desired PSF of all bands after homogenization. The required convolution kernel was calculated in each band by finding the kernel that minimizes the difference between the target PSF and the convolution product of this kernel with the current PSF. The images were then convolved with this kernel.

Then, a χ^2 detection image (Szalay et al., 1999), produced by combining NIR images of UltraVISTA ($YJHK_S$) with the optical z^{++} -band data from Subaru, was used to identify objects; the photometry for these objects in the other bands was obtained by running SExtractor in dual-image mode. Fluxes were extracted from 2'' to 3'' diameter apertures on PSF-homogenized images in each band, except for a few cases: for GALEX, the fluxes were measured using a PSF fitting method with the u^* -band image used as a prior, and for the SPLASH IRAC imaging, the IRACLEAN tool (Hsieh et al., 2012) was used to derive the photometry using the UltraVISTA $zYJHK_S \chi^2$ image as a prior. Photometry at 24 μm was obtained from the COSMOS MIPS-selected band-merged catalog (Le Floc'h et al., 2009). Far-IR photometry is provided by *Herschel* PACS (PEP guaranteed-time program, Lutz et al., 2011) and SPIRE (HERMES consortium, Oliver et al., 2012) at 100, 160, 250, 350, and 500 μm .

The Laigle et al. (2016) catalog provides secondary products as well, including photometric redshifts, stellar masses, and star formation rates. In particular, photometric redshifts were computed with *Le Phare* (Arnouts et al., 2002; Ilbert et al., 2006) with the same method as was used in Ilbert et al. (2013): the spectral energy distributions (SED) in 3 arcsec apertures were fit to a set of 31 templates, including spiral and elliptical galaxies from Polletta et al. (2007) and a set of 12 templates of young blue star-forming galaxies, produced using the models of Bruzual & Charlot (2003). Photometric redshifts for objects that are detected in X-rays were instead taken from Marchesi et al. (2016), who applied a set of templates that is more suitable for X-ray detected galaxies, in which the central black hole might significantly affect the UV/optical photometry (see section 2.1.3). Photometric redshift precision was characterized by comparing it with spectroscopic samples from the COSMOS spectroscopic master catalog (M. Salvato et al., in preparation).

Stellar masses were derived using *Le Phare*, with the same method as presented in Ilbert et al. (2015): a Chabrier (2003) initial mass function was assumed, and the photometry was fit with a library of synthetic spectra generated using the stellar population synthesis model of Bruzual & Charlot (2003), assuming both exponentially declining star formation histories (SFH) and delayed SFH ($\tau^{-2}te^{-t/\tau}$), assuming two different metallicities (solar and half-solar). Emission lines were added following the

prescription in Ilbert et al. (2009) together with two attenuation curves (the starburst curve of Calzetti et al. 2000 and a curve with a slope $\lambda^{0.9}$ from Appendix A of Arnouts et al. 2013). The E(B - V) values were allowed to vary in a range between 0 and 0.7. The masses were then obtained as the median of the marginalized probability distribution function (PDF).

2.1.2 IR data

We selected the starburst galaxies sample for this work from a FIR-selected catalog. In particular, we used the Gruppioni et al. (2013) catalog, which is selected from PACS/*Herschel* PEP observations in the COSMOS field. This catalog was also matched with the deep 24 μm imaging of Le Floc'h et al. (2009), with the HerMES extragalactic survey (Oliver et al., 2012) observed with SPIRE at 250, 350, and 500 μm in the same fields that were covered by PEP and with the IRAC-based catalog of Ilbert et al. (2010), including optical and NIR photometry and photometric redshifts.

Gruppioni et al. (2013) used as reference the blind catalogs at 100 and 160 μm from Berta et al. (2010, 2011), selected down to the 3σ level, which in COSMOS contain 5355 and 5105 sources at 100 and 160 μm , respectively. Then they associated their sources with the ancillary catalogs by means of a multiband likelihood ratio technique (Sutherland & Saunders, 1992; Ciliegi et al., 2001), starting from the longest available wavelength (160 μm , PACS) and progressively matching 100 μm (PACS) and 24 μm (MIPS). Their final catalog consists of 4110 and 4118 sources at 100 and 160 μm , respectively. Spectroscopic or photometric redshifts are available for 3817 and 3849 of their sources, respectively, which they used for the SED fitting in order to estimate the FIR luminosity function.

Stellar masses were obtained by fitting the broadband SED with a modified version of MAGPHYS (da Cunha, Charlot & Elbaz, 2008), which simultaneously fits the broadband UV-to-FIR observed SED of each object and ensures an energy balance between the absorbed UV light and the light that is reemitted in the FIR regime. The redshift for each object was fixed to the spectroscopic redshift when available or else to the photometric redshift; then the SED was fit to a best-fit model that we selected from a library built by combining different SFH, metallicities, and dust contents. Each SFH is the combination of an exponentially declining SFR model, to which random bursts of star formation are superimposed (see da Cunha et al., 2008, 2010). The emission of a possible AGN component was taken into account using a modified version

of the MAGPHYS code (SED3FIT, Berta et al., 2013), which adds a torus component to the modeled SED emission by combining the da Cunha et al. (2008) original code with the Fritz, Franceschini & Hatziminaoglou (2006) AGN torus library (see also Feltre et al., 2012).

In order to estimate the SFR of their sample, Gruppioni et al. (2013) calculated an SED using all the available multiwavelength data by performing a χ^2 fit using the *Le Phare* code (Arnouts et al., 2002; Ilbert et al., 2006) with the semiempirical template library of Polletta et al. (2007), which is representative of different classes of IR galaxies and AGN. They also added some modified templates in the FIR to better reproduce the observed Herschel data (see Gruppioni et al., 2010), and three starburst templates from Rieke et al. (2009). Then they integrated the best-fitting SED of each source over $8 \leq \lambda_{rest} \leq 1000 \mu\text{m}$ to derive the total IR luminosities ($L_{\text{IR}} = L[8\text{--}1000 \mu\text{m}]$) in 11 redshift bins (0.0–0.3, 0.3–0.45, 0.45–0.6, 0.6–0.8, 0.8–1.0, 1.0–1.2, 1.2–1.7, 1.7–2.0, 2.0–2.5, 2.5–3.0, and 3.0–4.2). Finally, they estimated the SFR_{IR} for all these sources from the total IR luminosity after subtracting the AGN contribution and using the Kennicutt (1998) relation (see also Gruppioni et al., 2015, for more details).

2.1.3 X-ray data

We used X-ray data from the *Chandra* COSMOS Legacy Survey (COSMOS-Legacy, Civano et al., 2016), a 4.6 Ms *Chandra* program that combines new observations obtained during *Chandra* Cycle 14 with the previous C-COSMOS Survey, allowing the X-ray data to uniformly cover the whole 2.2 deg² of the COSMOS field. The limiting fluxes are 2.2×10^{-16} , 1.5×10^{-15} , and $8.9 \times 10^{-16} \text{ erg cm}^{-2} \text{ s}^{-1}$ in the 0.5–2, 2–10, and 0.5–10 keV bands, respectively.

The optical counterparts to the X-ray COSMOS-Legacy sources are available in Marchesi et al. (2016). This was obtained by using the maximum likelihood ratio technique (e.g., Sutherland & Saunders, 1992; Brusa et al., 2005; Civano et al., 2012) by matching X-ray sources to three separate bands: i-band data from Ilbert et al. (2009), K_s -band data using UltraVISTA DR2, and IRAC 3.6 micron using either SPLASH or Sanders et al. (2007) sources from the Laigle et al. (2016) catalog in the UltraVISTA field. Optical counterparts were cross-correlated with the master spectroscopic catalog (Salvato et al. in prep), which contains spectroscopic redshifts from numerous observing campaigns and instruments. For the sources for which no spectroscopic redshift is available, a photometric redshift was provided that we obtained by follow-

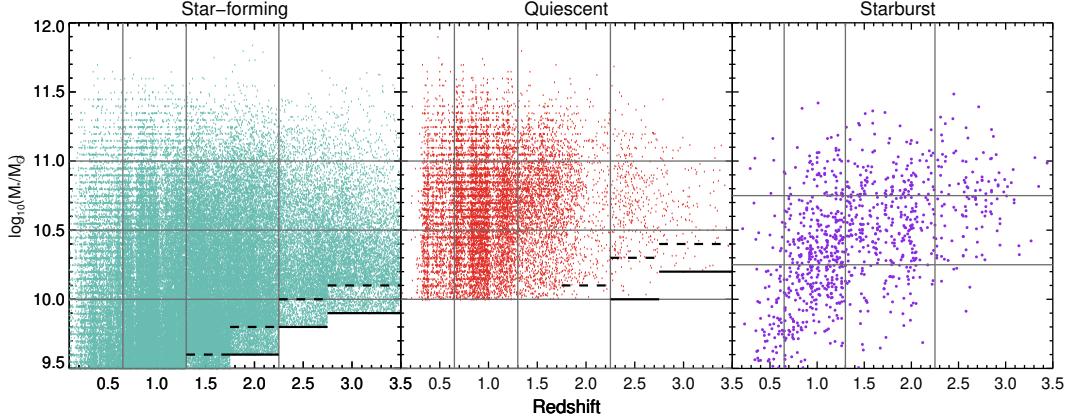


Figure 2.1: Stellar mass of our sample of galaxies shown as a function of redshift. Star-forming and quiescent galaxies are selected from Laigle et al. (2016) (green and red data points in the left and central panel, respectively), while the starbursts are selected from Gruppioni et al. (2013) (violet data points in the right panel). The solid and dashed black lines are the mass completeness thresholds are for UVista deep and ultradeep stripes from Laigle et al. (2016) for star-forming and quiescent galaxies. For starburst galaxies we considered as mass incomplete the low-mass bins ($9.50 \leq \log_{10}(M_\odot/M_*) \leq 10.25$) at $z > 1.30$. The solid gray horizontal and vertical lines show the limits of our mass and redshift bins.

ing the same procedure as in Salvato et al. (2011).

2.2 Sample selection

We selected a mass-complete sample from Laigle et al. (2016) by limiting our selection to the area that is covered by the UltraVISTA-DR2 observations (1.5 deg^2) in the redshift interval $0.1 < z < 3.5$. In order to separate star-forming from quiescent galaxies, we used the classification from Laigle et al. (2016), which is based on the rest-frame $\text{NUV}-r/r-J$ color-color diagram as in Ilbert et al. (2013). It allows separating dust-obscured galaxies from older stellar populations. In this diagram, galaxies with colors $\text{NUV}-r > 3(r-J) + 1$ and $\text{NUV}-r > 3.1$ are classified as quiescent.

We selected starburst galaxies from the Gruppioni et al. (2013) catalog in the same redshift interval down to $M_* \sim 10^{9.5} M_\odot$. Again, we chose this catalog because it is FIR selected, that is, selected based on the specific star formation rate (at least to a first approximation), and includes SFR_{IR} estimates, which allowed us to robustly estimate the SFR of highly star-forming galaxies that are heavily obscured. Based on Fig. 15 of Gruppioni et al. (2013), above $z \sim 1.3$ we probably miss the low-SFR starburst sources,

therefore we decided to show these data points as upper limits. The starburst galaxies were selected to have an SFR that is at least four times higher than the SFR of a typical main-sequence galaxy with a similar stellar mass and redshift, which is our definition of starburst galaxy and was previously introduced and motivated in Rodighiero et al. (2011). For the purpose of their selection, we considered the parameterization of the main-sequence from Schreiber et al. (2015) and obtained the relation

$$\begin{aligned} \text{SFR}_{IR}^{SB} &\geq 4 \times \text{SFR}_{MS}(z_{ave}, M_*) = \\ &= 4 \times [m - m_0 + a_0 r - a_1 [\max(0, m - m_1 - a_2 r)]^2], \end{aligned} \quad (2.1)$$

where SFR_{IR}^{SB} is the star formation rate of a starburst galaxy obtained from its IR luminosity, $r = \log_{10}(1+z)$, $m = \log_{10}(M_*/10^9 M_\odot)$, and the parameters $m_0 = 0.5$, $m_1 = 0.36$, $a_0 = 1.5$, and $a_1 = 0.3$. We used SFR_{IR}^{SB} , M_* , and z from the Gruppioni et al. (2013) catalog. We matched the starburst sample with the catalog of star-forming and passive galaxies that was previously color-color selected from Laigle et al. (2016), using a $3.5''$ association radius in order to exclude the starburst galaxies from our star-forming and quiescent selection. In order to reduce the chances of including spurious sources in our cross matches, we only accepted matches that have a $\Delta z = |z_{\text{Gruppioni}} - z_{\text{Laigle}}| < 0.3$ for a distance greater than $2''$, allowing us to include seven more starburst galaxies in our sample. We compared the stellar masses of the starburst sample (derived from MAGPHYS) with their stellar masses from Laigle et al. (2016) (derived from *Le Phare*) in each of the redshift bins and found that they lie along the 1:1 relation with increasing scatter with redshift (RMS ranging from 0.18 dex at low z to 0.35 dex at high- z), but with no offset (the median of the difference of the masses is about 10^{-2}). From the cross-match we find that 965 color-selected star-forming galaxies are IR-classified as starbursts (this is expected because the $\text{NUV}-r/r-J$ diagram does not have a starburst area) and 28 color-selected quiescents are IR starbursts. We moved 1494 galaxies that are $24\mu\text{m}$ MIPS/*Spitzer* detected from our quiescent selection to the star-forming galaxy sample because the IR emission indicates that a certain amount of optically hidden star formation is ongoing. For the X-ray detected star-forming and quiescent galaxies we used the redshift included in Marchesi et al. (2016). We have a final number of 83,904 star-forming, 12,839 quiescent, and 1,003 starburst galaxies.

We further divided the sample into four redshift bins: $0.1 < z < 0.65$, $0.65 < z < 1.3$, $1.3 < z < 2.25$, and $2.25 < z < 3.5$.

Every redshift bin was chosen with the purpose of including a sufficient num-

ber of galaxies (see Table 2.1), and then the galaxy sample of each redshift bin was divided into mass bins. Our final sample is shown in Fig. 2.1 together with the mass-completeness thresholds from Laigle et al. (2016) for star-forming and quiescent galaxies and the limits of our stellar mass and redshift bins.

2.3 Method

The aim of this paper is to study the average BHAR for galaxies with different star formation activities, from passive galaxies to starbursts, at different cosmic epochs and at different stellar masses. Because AGN activity is a stochastic event, as the AGN duty cycle is short compared to the star formation episodes in galaxies, it is unlikely that a galaxy has been observed at the peak of its black hole accretion. This means that if we limited the analysis to bright X-ray galaxies, we would obtain a biased view of the average AGN activity in galaxies. For this reason, we complement X-ray individual detections with the stacking analysis on *Chandra* X-ray images for nondetected galaxies.

2.3.1 X-ray stacking analysis

In order to estimate the average X-ray luminosity for each mass and redshift subsample, we followed the same method as in Rodighiero et al. (2015). We used individual X-ray detections from the COSMOS-Legacy catalog (Civano et al., 2016) when possible, taking advantage of the match between optical and X-ray sources performed by Marchesi et al. (2016) (see section 2.1.3). For the non-X-ray detected sources we performed stacking on the same images using the CSTACK tool v4.32¹ (Miyaji et al., 2008).

We propagated the probability distributions of count rates with Monte Carlo simulations. We focused on the 2-7 keV band and took advantage of the CSTACK bootstrap output. We simulated it 10^6 times by interpolating the inverted cumulative distribution function of the bootstrap probability distribution. For X-ray detections, we instead assumed a Gaussian probability distribution and a σ equal to the error on the count rate from the catalog when available. Alternatively, we set the error on the detection to $\sigma = 0.25 \times CR(2 - 7 \text{ keV})$. Count-rate distributions from the stacking

¹<http://cstack.ucsd.edu/cstack> or <http://lambic.astrosen.unam.mx/cstack/> developed by Takamitsu Miyaji.

Table 2.1: Number of X-ray (2-7 keV band) stacked and detected galaxies per redshift bin and galaxy type.

Redshift bins	Star-forming		Quiescent		Starburst	
	Detections	Stacked	Detections	Stacked	Detections	Stacked
$0.10 < z < 0.65$	249	8,211	45	1,968	12	66
$0.65 < z < 1.30$	743	28,037	116	7,055	22	366
$1.30 < z < 2.25$	764	31,012	42	2,835	48	280
$2.25 < z < 3.50$	337	11,615	16	443	9	115

and detections were converted into standard flux in the 2 – 10 keV band through the WebPIMMS² conversion factor obtained for *Chandra* cycle 14 by assuming a power-law spectrum for the AGN with photon index $\Gamma = 1.8$ and galactic hydrogen absorption ($N_H = 2.6 \times 10^{20} \text{ cm}^{-2}$, Kalberla et al., 2005). This value of the photon index Γ is compatible with the cosmic X-ray background according to Ananna et al. (2019) for cutoff energies $E_c < 100 \text{ keV}$. We converted fluxes into rest-frame with a K-correction with the shape of $K_{corr} = (1+z)^{\Gamma-2}$, where z is the average redshift of each subsample and Γ is the photon index just introduced.

Because we search for a global average X-ray measurement by combining detections and non-detections in each redshift and mass bin, we applied the following equation to all the fluxes in the probability distribution of each detection and the stacking:

$$F_{\text{ave}} = \frac{\sum_{j=1}^{n_{\text{detected}}} F_{j,\text{detected}} + n_{\text{stacked}} \cdot F_{\text{stacked}}}{n_{\text{detected}} + n_{\text{stacked}}}, \quad (2.2)$$

where in each redshift and mass bin, n_{detected} is the number of detections, $F_{j,\text{detected}}$ is the flux of each detection, n_{stacked} is the number of undetected objects that were therefore stacked, and F_{stacked} is the total flux from the stacked objects. This returns a probability distribution of rest-frame flux for each redshift and mass bin. We show in Table 2.1 the number of stacked and detected galaxies.

2.3.2 X-ray luminosity and black hole accretion rate estimate

We converted the distribution of average rest-frame fluxes F_{ave} (Equation 2.2) to average luminosities L_{ave} using the corresponding luminosity distance at the average redshift of the sources included in the bin. We estimated and subtracted the contribution from the stars (young and old stellar populations) in the 2-10 keV range as in

²<http://cxc.harvard.edu/toolkit/pimms.jsp>

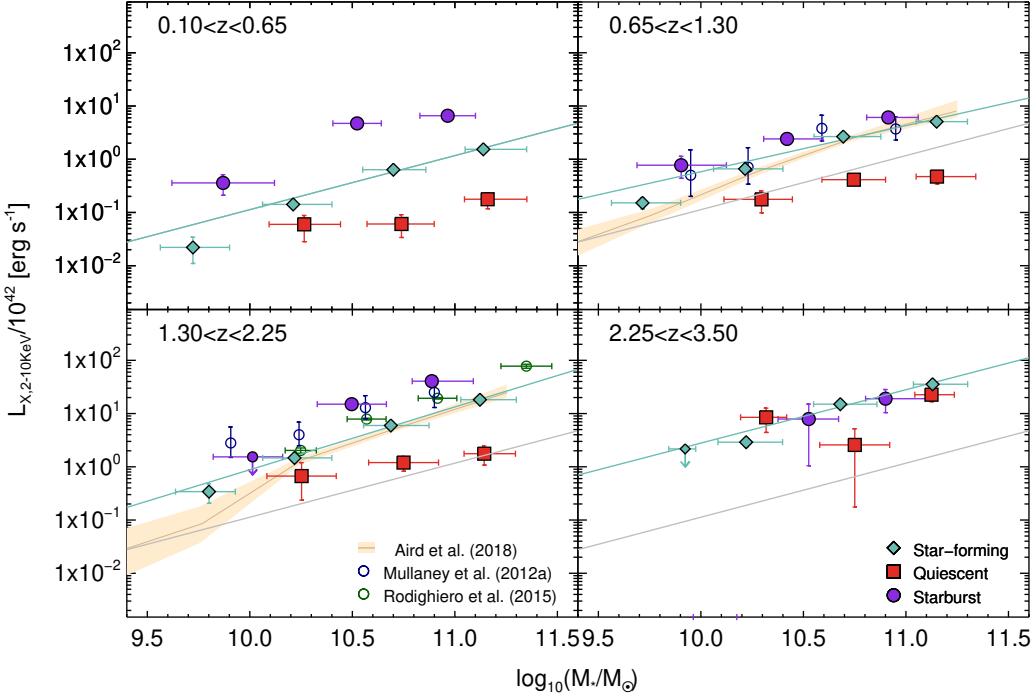


Figure 2.2: X-ray luminosity in the 2-10 keV band for our sample as a function of stellar mass. Each panel represents one of the four redshift bins that we considered. The light green diamonds represent the normal star-forming sample, the red squares represent the quiescent sample, and the violet circles represent the starburst sample. The data points represent the median value of the stellar mass and the X-ray luminosity. Vertical error bars represent the 5th and 95th percentile (90% confidence interval) of the distributions, while horizontal error bars correspond to 1σ . We show an upper limit (the 95th percentile) for the data points whose X-ray luminosity 5th percentile is compatible with zero and for starburst galaxies at high redshift and low mass, where our sample is incomplete. The light green continuous line is the best fit for the star-forming galaxy data points at each redshift, while the gray line shows the best fit in the lowest redshift bin, shown for comparison purposes. We also report data points from Mullaney et al. (2012a) at an average redshift of $z = 1, 2$ (open dark blue circles, Rodighiero et al. (2015) at an average redshift of $z = 2$ (open green circles) and Aird et al. (2018) at $z = 1, 2$ (solid brown curve; the yellow area shows the 1σ confidence interval). The data points from Rodighiero et al. (2015) and Aird et al. (2018) are scaled to the same k -correction as we adopted here.

Table 2.2: X-ray luminosity fits to the equation $\log_{10} L_X = a \log_{10} M_* + b$ for the star-forming galaxies of our sample. Errors are 1σ uncertainty estimates of parameters

Redshift bin	a	b
$0.10 < z < 0.65$	1.02 ± 0.03	-11.1 ± 0.3
$0.65 < z < 1.30$	0.87 ± 0.02	-8.9 ± 0.2
$1.30 < z < 2.25$	1.18 ± 0.03	-11.8 ± 0.3
$2.25 < z < 3.50$	1.00 ± 0.05	-9.6 ± 0.5

Lehmer et al. (2016) using their best-fit values,

$$L_{2-10\text{ keV}}(M_*, z, \text{SFR}) [\text{erg s}^{-1}] = \alpha_0(1+z)^\gamma M_* + \beta_0(1+z)^\delta \text{SFR}, \quad (2.3)$$

with $\log \alpha_0 = 29.37$, $\log \beta_0 = 39.28$, $\gamma = 2.03$, and $\delta = 1.31$. This correction has two terms that take into account the X-ray emission from the young stellar populations, that is, high-mass X-ray binaries, whose emission is proportional to the SFR of the galaxy (see Sec. 2.3.3), and from the old stellar populations, that is, low-mass X-ray binaries, whose emission is proportional to the stellar mass of the galaxy. This is particularly important in quiescent galaxies. We confirmed the effect of this correction by comparing our results with the corrections from Fornasini et al. (2018) and Aird et al. (2017). The correction from Aird et al. (2017) gives no apparent difference to our results, while the correction from Fornasini et al., which is to be considered an upper limit because it includes some residual AGN emission, has some effect at the lowest redshift that causes some data points to become compatible with zero and a lower normalization by ~ 0.2 dex in star-forming and quiescent galaxies. The overall results are not affected by it, however.

In order to correct for the average internal absorption from the galaxy (N_H), we estimated the hardness ratios (HR) of our sample. We define the HR as

$$\text{HR} = \frac{H - S}{S + H}, \quad (2.4)$$

where H and S are the median of the distributions of count rates in the 2-7 keV and 0.5-2 keV band, respectively. We compared the estimated HRs of our data with those obtained from models in PyXspec. The model we used is PHABS*(CABS*ZPHABS*PO+APEC), which is composed of a redshifted and absorbed power-law emission with slope $\Gamma = 1.8$ and hot ionized gas emission at $kT = 1$ keV. The model also includes galactic absorption. The power-law spectrum was normalized because the fluxes we measured and the APEC warm gas component

were normalized to equal the emission expected from the SFR of the galaxies (Mineo et al., 2012, Eq. 1). Finally, we convolved the model with an auxiliary response file (ARF) from *Chandra* ACIS-I cycle 14 in order to obtain the *H* and *S* count rates the instrument would have measured. In fact, the instrument measures a count rate that is a function of the energy spectrum and the effective area which is a function of the photon energy, so the convolution is necessary to obtain consistent measures between model and observations. All models give two N_H values at which the HRs are compatible with our data. In fact as the N_H increases it will absorb the soft emission first, thus making the spectrum harder and producing a first match with the measured hardness ratio. Then, when the N_H has absorbed most of the AGN power law emission, the gas' soft emission begins to dominate, thus producing a second hardness ratio match with the data. The only exception having just one N_H solution are star-forming galaxies in the lowest redshift bin, where the modeled spectra are too soft, probably because the 1 keV gas emission is overestimated. In the remaining cases, we find a possible solution at $N_H = 10^{22.0-23.2} \text{ cm}^{-2}$ and another at higher N_H values $N_H = 10^{23.5-24.8} \text{ cm}^{-2}$, where all the power-law emission has been absorbed and only the hot gas is visible. The two N_H solutions increase with redshift. We decided to use the solution with a lower N_H because we cannot observe galaxies with Compton-thick levels of obscuration. Even though we performed stacking, the number of photons from heavily obscured AGN in the energy range of *Chandra* is very low and does not dominate our population, since $\sim 99\%$ of photons are absorbed in the 0.2-8keV *Chandra* band when $N_H \simeq 10^{24}$. Furthermore, our NIR selection may already exclude dust-obscured galaxies, which are thought to be heavily absorbed in the X-rays (Fiore et al., 2008; Corral et al., 2016; Riguccini et al., 2019). A last consideration is that even though many works reported an intrinsic fraction of highly obscured AGN or Compton-thick AGN of about $\sim 30 - 50\%$ at various redshifts (Ueda et al., 2014; Vito et al., 2014; Ricci et al., 2015; Ananna et al., 2019), it still is not possible to detect these galaxies in the *Chandra* X-ray energies. The HRs of our sample do not show a trend with mass, therefore we decided to use the same obscuration correction for all mass bins of a given galaxy type and redshift. We estimated the correction factors using WebPIMMS, and it extends from absorbed to unabsorbed flux in the 2-10 keV band at the average redshift of the bin and with the N_H found from the HR analysis for each galaxy type. These corrections are approximately 5 – 10%.

Finally, we considered the median value of the X-ray luminosity distribution as the representative luminosity of each bin and the 5th and 95th percentiles as the lower and

upper limits to the uncertainty associated with the X-ray luminosity. We considered values compatible with zero when the 5th percentile assumed a negative value, and in these cases, we show them as upper limits.

We transformed the X-ray binaries and obscuration-corrected X-ray luminosity into a BHAR as in Merloni & Heinz (2008), Mullaney et al. (2012a), Delvecchio et al. (2014), Rodighiero et al. (2015), and Baronchelli et al. (2018) with the relation

$$\text{BHAR}(M_*, z) = \frac{(1 - \epsilon) \times L_{\text{bol}}(M_*, z)}{\epsilon c^2}, \quad (2.5)$$

where L_{bol} is the AGN bolometric luminosity obtained using a luminosity-dependent bolometric correction consisting of (i) the bolometric correction used in Yang et al. (2018), which is a modified version of the bolometric correction from Lusso et al. (2012), down to $L_X = 10^{42.4} \text{ erg s}^{-1}$, and (ii) $k_{\text{bol}} = 16$ for lower luminosities as in She et al. (2017). c is the speed of light in vacuum, and ϵ is the efficiency by which mass is converted into radiated energy in the accretion process. Here we assumed $\epsilon = 0.1$ (e.g., Marconi et al., 2004; Mullaney et al., 2012a; Rodighiero et al., 2015; Baronchelli et al., 2018), or that roughly 10% of the accreted rest-mass is converted into radiant energy, regardless of M_{BH} .

2.3.3 Star formation rates: FIR stacking and UV SED fitting

For star-forming and quiescent galaxies, we estimated the SFR as the sum of the IR and far-UV-observed contributions ($SFR_{\text{IR+UV}}$), while for starburst galaxies, we only considered the SFR_{IR} . The SFR_{IR} was derived through the empirical calibration of Kennicutt (1998, Eq. 4) from the total IR luminosity (L_{IR}). The SFR_{UV} (not corrected for extinction) was inferred, following the prescription of Kennicutt (1998, Eq.1), from the rest-frame luminosity at 1600 Å ($L_{1600\text{\AA}}$). The two SFRs were then combined as discussed by Nordon et al. (2013), and they were finally converted into the Chabrier (2003) IMF as in Cimatti et al. (2008) by dividing by a factor of 1.7,

$$SFR_{\text{IR+UV}} [M_{\odot} \text{ yr}^{-1}] = SFR_{\text{UV}} + SFR_{\text{IR}} = (2.86 L_{1600\text{\AA}} + 1.7 L_{\text{IR}}) \times \frac{10^{-10}}{1.7} [L_{\odot}]. \quad (2.6)$$

In order to obtain the $SFR_{\text{IR+UV}}$ for star-forming and quiescent galaxies, we performed bootstrapping, in which we selected a subsample of galaxies in a M_* and z bin at each iteration and estimated their L_{IR} and $L_{1600\text{\AA}}$. We then combined the median of each luminosity through Eq. 2.6 and thus derived the $SFR_{\text{IR+UV}}$ distribution. L_{IR} was obtained by stacking 160 μm PACS/*Herschel* maps as in Rodighiero et al. (2014). In

2.4. COMPARING THE SMBH X-RAY EMISSION AT THE DIFFERENT GALAXY LIFE STAGES THROUGHOUT COSMIC TIME

our procedure we accounted for detections and nondetections to obtain the final median stacked fluxes that we then converted into bolometric luminosities by adopting an average k -correction (Chary & Elbaz, 2001). Because the $L_{1600\text{\AA}}$ luminosity is not included in Laigle et al. (2016), we derived it in the following way. For each source in the sample, we reconstructed the galaxy SED using *Hyperzmass*, which is a modified version of the *Hyperz* software (Bolzonella et al., 2000, 2010) and is suitable for computing the stellar mass when the photometric redshift is known. We adopted the Bruzual & Charlot (2003) stellar population models, with exponentially declining SFH, that is, $\text{SFR} \propto e^{t/\tau}$, with $\tau=0.1, 0.3, 1, 2, 3, 5, 10, 15, 30, \infty=\text{constant SFR}$. The $L_{1600\text{\AA}}$ were extracted directly from the *Hyperzmass* best-fit templates. The SFRs for quiescent and star-forming galaxies are shown in the upper left and central panels of Fig. 2.3 as a function of stellar mass. They are color-coded based on redshift.

For starburst galaxies we used the SFR_{IR} from Gruppioni et al. (2013) (see Section 2.1.2) after verifying that their SFR_{UV} is negligible. Similarly to star-forming and quiescent galaxies, we performed bootstrapping for every M_* and z bin, where in each loop we derived a median SFR_{IR} and therefore an SFR_{IR} distribution. The SFRs for starburst galaxies are shown in the top right panel of Fig. 2.3.

2.4 Comparing the SMBH X-ray emission at the different galaxy life stages throughout cosmic time

In Figure 2.2 we show the X-ray 2-10 keV luminosity (L_X) from Equation 2.3 as a function of stellar mass for our sample, divided into different redshift bins. For each bin we compared the X-ray luminosity for the three types of galaxies: star-forming, quiescent, and starbursts. Our data points are centered on the median X-ray luminosity and stellar mass. Vertical error bars are the 90% confidence range. We show a log-log linear fit for star-forming galaxy data points. The fits were performed ignoring the upper limit data points.

We find a robust relation between the average L_X and M_* in star-forming galaxies: this indicates that black holes in more massive galaxies grow faster than black holes in the less massive galaxies. This might in part be sustained by the higher fraction of type I AGN at high stellar masses, as seen by Suh et al. (2019) for *Chandra* COSMOS Legacy detected sources. Moreover, this relation increases in normalization by about

1.5 dex but maintains an almost constant slope of about ~ 1.00 toward higher redshift, which indicates that black holes grew faster at earlier epochs than they do today. This is shown in Table 2.2, which lists the best linear fit to our data ($\log L_X = a \log M_* + b$). This relation looks very similar to the main-sequence of star-forming galaxies, which is the relation between SFR and M_* , as we show more clearly in Figure 2.3. The quiescent $24\mu\text{m}$ MIPS/*Spitzer* detected sources are a small percentage of the overall star-forming sample ($\sim 2\%$), but they are mostly concentrated at the high masses and low redshifts, where they reach a percentage as high as 86% at $M_* > 10^{11} M_\odot$ in the lowest redshift. The addition of these sources had very little effect on the L_X of the higher mass bins, but all bins show a systematical increase in L_X of $\lesssim 0.1$ dex.

Starburst galaxies show average L_X with a similar dependence on stellar mass as star-forming galaxies and a mild dependence on redshift: as a result, the X-ray luminosities of starbursts are $\sim 0.4 - 1.1$ dex higher than those of star-forming galaxies at $z \simeq 0.4$, but the luminosities at $z \simeq 2 - 3$ are compatible. If these galaxies undergo a major merger event, as is commonly believed, this suggests that while a higher availability of cold gas allows the SFR to increase considerably with respect to a galaxy in a secular evolution phase, this gas availability is not able to accrete onto the black hole at a pace higher than a certain threshold that does not vary as much as the star-forming population (~ 0.5 dex) in the redshift range we covered.

Finally, quiescent galaxies have X-ray luminosities that tend to be lower than those of star-forming galaxies. They vary with increasing mass from ~ 0.5 dex to ~ 1 dex in difference, with the exception of the highest redshift bin, which is almost compatible with the luminosity of main-sequence galaxies. The resulting relation with stellar mass is flatter than the relation observed for star-forming galaxies. The relation is limited to just a few (high-) mass bins because in the mass regime below $10^{10} M_\odot$, the X-ray flux was not high enough to constrain the X-ray luminosity of the subsample: we obtained a 95th percentile luminosity value < 0 . According to the hardness ratio analysis performed by Paggi et al. (2016) on a subsample from C-COSMOS of early-type stacked galaxies, the X-ray luminosities of our quiescent galaxies are expected to be compatible with a combination of thermal and AGN emission in the lower two redshift bins and with highly obscured AGN at higher redshifts. This confirms the origin of the emission.

In Figure 2.2 we compare our results for the sample of star-forming galaxies with literature results by Mullaney et al. (2012a), Rodighiero et al. (2015), and Aird et al. (2018) after scaling the luminosity to the same k -correction as we used in our analysis

Table 2.3: Fit parameters of the relations in Fig. 2.3. The equations we used are for $\log_{10} \text{SFR} = m \log_{10} M_* + q$; for $\log_{10} \text{BHAR} = m \log_{10} M_* + q$; and for the ratio $\log \frac{\text{BHAR}}{\text{SFR}} = \alpha \log M_* + \beta$.

Redshift bin	SFR		BHAR		BHAR/SFR	
	m	q	m	q	α	β
Star-forming						
$0.10 < z < 0.65$	0.71 ± 0.02	-6.4 ± 0.2	1.02 ± 0.03	-13.7 ± 0.3	0.72 ± 0.09	-11.6 ± 0.9
$0.65 < z < 1.30$	0.649 ± 0.013	-5.49 ± 0.13	0.87 ± 0.02	-11.5 ± 0.2	0.58 ± 0.03	-9.9 ± 0.4
$1.30 < z < 2.25$	0.639 ± 0.018	-5.06 ± 0.19	1.18 ± 0.03	-14.4 ± 0.3	0.65 ± 0.04	-10.6 ± 0.4
$2.25 < z < 3.50$	0.60 ± 0.04	-4.3 ± 0.4	1.00 ± 0.05	-12.2 ± 0.5	0.53 ± 0.07	-9.3 ± 0.7
Quiescent						
$0.10 < z < 0.65$	0.40 ± 0.14	-4.5 ± 1.5	0.58 ± 0.17	-9.9 ± 1.9	-	-3.38 ± 0.07
$0.65 < z < 1.30$	0.37 ± 0.15	-3.6 ± 1.6	0.41 ± 0.16	-7.4 ± 1.7		
$1.30 < z < 2.25$	-0.2 ± 0.5	3.1 ± 4.9	0.5 ± 0.2	-7.6 ± 2.4		
$2.25 < z < 3.50$	-	-	0.59 ± 0.18	-7.9 ± 2.0		
Starburst						
$0.10 < z < 0.65$	0.73 ± 0.05	-5.8 ± 0.5	0.86 ± 0.09	-11.0 ± 1.0	0.36 ± 0.12	-7.9 ± 1.3
$0.65 < z < 1.30$	0.61 ± 0.03	-4.3 ± 0.3	0.86 ± 0.10	-11.1 ± 1.1	0.22 ± 0.11	-6.7 ± 1.2
$1.30 < z < 2.25$	0.54 ± 0.07	-3.1 ± 0.8	1.08 ± 0.11	-12.8 ± 1.2	0.54 ± 0.15	-9.6 ± 1.6
$2.25 < z < 3.50$	0.59 ± 0.14	-3.4 ± 1.5	1.0 ± 0.7	-12.5 ± 7.3	0.44 ± 0.72	-9.2 ± 7.8

where necessary (i.e., rescaled to the same photon index Γ in the k -correction). At $z \sim 1$, our data points agree well with those of Mullaney et al. (2012a) and Aird et al. (2018). At $z \sim 2$ our results are in great agreement, within the errors, with the results of Aird et al. (2018), while our X-ray luminosities are systematically lower than those by Mullaney et al. (2012a) and Rodighiero et al. (2015).

2.5 Constraining the coevolution of galaxy and black hole accretion

In Figure 2.3 we compare the SFR and the BHAR as a function of M_* for the three categories of galaxies. In the left column we show star-forming galaxies, in the middle column we list quiescent galaxies, and the right column contains starburst galaxies. The top panel shows the total $\text{SFR}_{\text{UV+IR}}$ (SFR_{IR} only for starburst galaxies, see Sec. 2.3.3) for all the redshift bins, the middle panel shows the BHAR of the sample

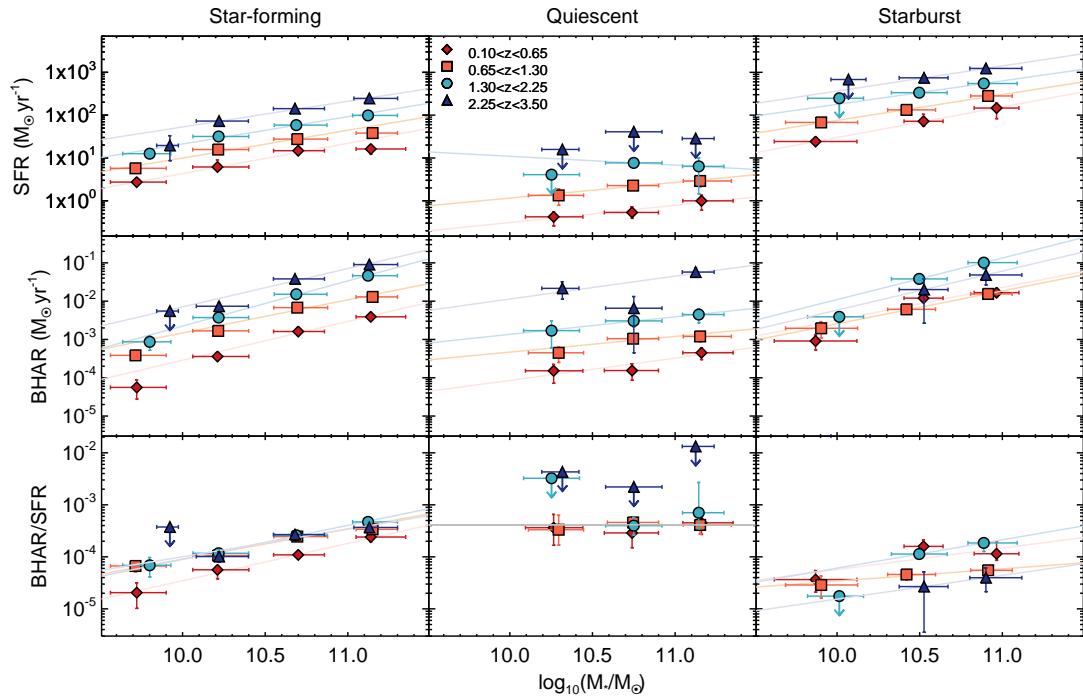


Figure 2.3: Comparison between the average SFR and the average BHAR for the three samples: normal star-forming galaxies (left), quiescent galaxies (center), and starburst galaxies (right). In all plots data points are color- and shape-coded according to the redshift interval. The data points are the median value, and vertical errors represent the 5th and 95th percentile. We show an upper limit (the 95th percentile) for the data points whose SFR or BHAR 5th percentile is compatible with zero and for starburst galaxies at high redshift and low mass, where our sample is incomplete in mass. The error bars shown on the stellar masses correspond to 1σ of the distribution. The solid light colored lines are the best fits and are color-coded according to each redshift bin (see Table 2.3). Top panel: M_* -SFR relation. Middle panel: M_* -BHAR relation. Bottom panel: BHAR-to-SFR ratio as a function of stellar mass. The dotted gray line for quiescent galaxies is at a constant value of $\text{BHAR/SFR} = 3.8 \times 10^{-4}$.

obtained from the X-ray luminosity (Figure 2.2) using Equation 2.5, and the bottom panel presents the ratio between BHAR and SFR. We show linear fits in all panels except for the BHAR/SFR of quiescent galaxies where we report the average value of all redshifts and mass bins because no evolutionary trends are apparently visible. The best-fit parameters are shown in Table 2.3.

2.5.1 Star-forming galaxies

For star-forming galaxies, the top panel shows the well-known evolution of the main-sequence of star-forming galaxies with time: the SFR follows a sublinear relation (in logarithmic units) with stellar mass that evolves in normalization with redshift, maintaining an almost constant slope and a bend at high M_* in the two lower redshift bins. This bend was introduced by including the quiescent $24\mu\text{m}$ MIPS/*Spitzer* detected sources, which as described above, are concentrated at high masses and lower redshifts. The median SFR_{UV} of the sample is lower than the median SFR_{IR} by a factor that varies from 3 (low mass) to 100 (high mass, see Table A.1), which means that the SFR_{UV} is negligible in most cases.

The observed BHAR distribution has a higher slope in the log-log plane; it is slightly superlinear (a direct consequence of the $L_X - M_*$ relation). It is present at all considered redshifts and evolves with it. We note that the decrease in BHAR normalization evolves faster in the lowest redshift bin, but maintains a more constant increase at earlier cosmic epochs. The total decrease in BHAR at given M_* is about 1.5 dex, and in SFR, the decrease is about 1-1.2 dex.

The ratio of BHAR to SFR provides interesting insights into how these two phenomena relate to each other. It increases with mass with a slope that varies in the $\sim 0.5 - 0.7$ range, decreases at higher redshifts, and does not evolve in normalization. The positive slope of the BHAR/SFR can be interpreted as an indication of an increased accretion efficiency in galaxies with high stellar mass by driving the infalling gas directly toward their inner core, which fuels a faster accretion onto the black hole. This can also be considered to mean that the main driver of the black hole accretion is not time, but the initial stellar mass. We note that the analogy between the increase in the BHAR/SFR as a function of stellar mass and the increase in the density of the central kiloparsec of the galaxy Σ_1 with mass (Fang et al., 2013; Barro et al., 2017). It has been proposed that more massive galaxies might have denser cores that might allow for a faster growth of the black hole.

2.5.2 Quiescent galaxies

In the central column of Fig. 2.3 we show the data points of the quiescent galaxies. As expected, the SFR of these galaxies is lower than that of star-forming galaxies. The two lower redshifts show a weak dependence of stellar mass, with a slope of ~ 0.4 , and the evolution with redshift in the bins in which we were able to constrain the SFR is also clear. The reason might be that high-redshift elliptical galaxies are of a different nature than local galaxies and have higher percentages of gas (Gobat et al., 2018), but part of the emission might also originate in dust cirrus illuminated by older stellar populations (Draine & Li, 2007; Béthermin et al., 2015). Another factor that may be contributing to the evolution we see is a small fraction of misclassified star-forming galaxies at high redshift. This may be caused by galaxies that emit in the MIPS band but lie below the detection threshold or by cross-contamination in the $\text{NUV} - r/r - J$ because the uncertainty in the estimation of the rest-frame magnitudes that are required to classify a galaxy in the color-color diagram is higher. Overall, it is not possible to determine whether this evolution in time and mass dependence of quiescent galaxies and their mass dependence is significant, especially because we are able to constrain only these high-mass bins.

For comparison with star-forming galaxies, we note that the median SFR_{UV} of the quiescent sample is lower by a factor of about 15 than the median SFR_{IR} (see Table A.2). This means that the SFR_{UV} is negligible in most cases.

We are better able to constrain the BHAR at all redshifts and see a redshift evolution and weak dependence on stellar mass, with a slope ~ 0.5 , even though it is difficult to analyze the slope evolution because of the small number of data points. The normalization in BHAR is higher than in SFR, that is, the normalization of the BHAR is similar to the normalization of BHAR in star-forming galaxies, while the SFR of quiescents is clearly lower than that of star-forming galaxies. This indicates that the efficiency in accreting material onto the black hole is higher than in forming stars in galaxies at this late stage in the life of galaxies: the gas present in the galaxy “prefers” to fall into the black hole rather than form stars. This is consistent with the results of Gobat et al. (2018), who reported that a substantial amount of gas is available at high redshift in quiescent galaxies. This is then probably consumed less efficiently than in star-forming galaxies.

The BHAR/SFR of quiescent galaxies, for the few mass and redshift bins where it was possible to constrain this, shows a flat trend in mass that is compatible with a

constant value of 4.2×10^{-4} at all redshifts and M_* . This value, obtained as a weighted mean of the data points, is a confirmation of the trends we have seen for BHAR and SFR: it is compatible with the values obtained for the highest mass bin of star-forming galaxies, indicating that even though the ability of a galaxy to form stars has decreased in these galaxies, the efficiency of attracting the gas to the black hole did not.

2.5.3 Starburst galaxies

In the right column of Figure 2.3 we show the starburst sample. By selection, FIR-selected starbursts have higher SFR than normal star-forming galaxies; the evolution in time is clear, as is the dependence on M_* .

In starburst galaxies, the BHAR shows a close to linear dependence on mass and a weak evolution in redshift: values range in about 1 dex, but the highest redshift bin has a lower normalization than the previous one, differently from what is seen in the other phases of galaxy life. This decrease in BHAR, which is not as pronounced as the decrease in star-forming galaxies (about ~ 1.5 dex) causes them to slowly become starburst galaxies (i.e., higher than main-sequence galaxies) in the black hole accretion as well at lower redshift.

There is no clear evolution in the normalization of the BHAR/SFR, but it is interesting to note that the highest redshift bin shows the lowest normalization, which indicates that in these extremely star-forming galaxies black hole accretion was disfavored at higher redshifts. The slope appears to be positive and of about $0.2 - 0.4$.

We tested selecting a starburst galaxy sample using the main-sequence from Rodighiero et al. (2011) with a redshift evolution as in Sargent et al. (2012), $SFR_{MS}(z) \propto (1+z)^{2.8}$. This selection assumed no bend on the main-sequence at high stellar masses and was calibrated up to $z \sim 2$. When we extrapolated this evolution to higher redshifts, we found a higher normalization than was reported by Schreiber et al. (2015), which instead was calibrated up to $z \sim 5$. This starburst selection led to no significant differences in the SFR of the sample, but to slightly lower BHAR values. We were also unable to constrain the X-ray luminosities of starbursts in the highest redshift bin. This suggests that the most starbursty galaxies accrete their black hole even less. This finally resulted in a flatter trend of the BHAR/SFR against stellar mass.

2.5.4 Comparisons with the literature

In Figure 2.4 we compare our results for the star-forming sample with results from the literature. We show our data points from Figure 2.3 in the left column and compare them with results from Yang et al. (2018) as solid lines, with results from Aird et al. (2019) as a dashed beige line, and with results from Delvecchio et al. (2019) as a dot-dashed violet line.

Our data points qualitatively agree with those from Yang et al. (2018) in the three panels. Yang et al. (2018) selected their star-forming sample from GOODS-N, GOODS-S, and COSMOS UltraVISTA DR1 based on the SFR from SED fitting with a SFR threshold at 1.3 dex below the main-sequence. SFRs are from Santini et al. (2015) and Barro et al. (2019)). Their average SFRs are slightly higher than ours. They also reported that the BHAR of the lowest redshift bin is separated more than the BHAR between the other bins.

The BHAR/SFR of Aird et al. (2019) has a linear trend with mass with a higher slope than ours. Their BHARs were obtained with the same method as in Aird et al. (2018). Their SFRs are from SED fitting from the UV to MIR, which means that the FIR was not included. Together with the constant bolometric correction they used, this might lead to the higher slope value.

We also compared our data points with the relation found by Delvecchio et al. (2019) through an empirically motivated model that successfully reproduces the observed X-ray luminosity function (XLF) since $z \sim 3$. With this model, they found a growth in two steps: until the galaxy reaches a critical mass, the black hole growth lags behind it, and then, as the stellar mass increases, the BHAR is enhanced with respect to the SFR, following a superlinear relation very similar to ours, except for an offset of $\sim 0.1 - 0.3$ dex.

2.6 Comparison between the evolution of sBHAR and sSFR

We computed the average sSFR (i.e., specific SFR, SFR/M_*) and the sBHAR (i.e., specific BHAR, $\text{BHAR}/\text{M}_{\text{BH}}$) for all the bins in M_* and z . In this section we describe how the sBHAR was derived and we discuss the results.

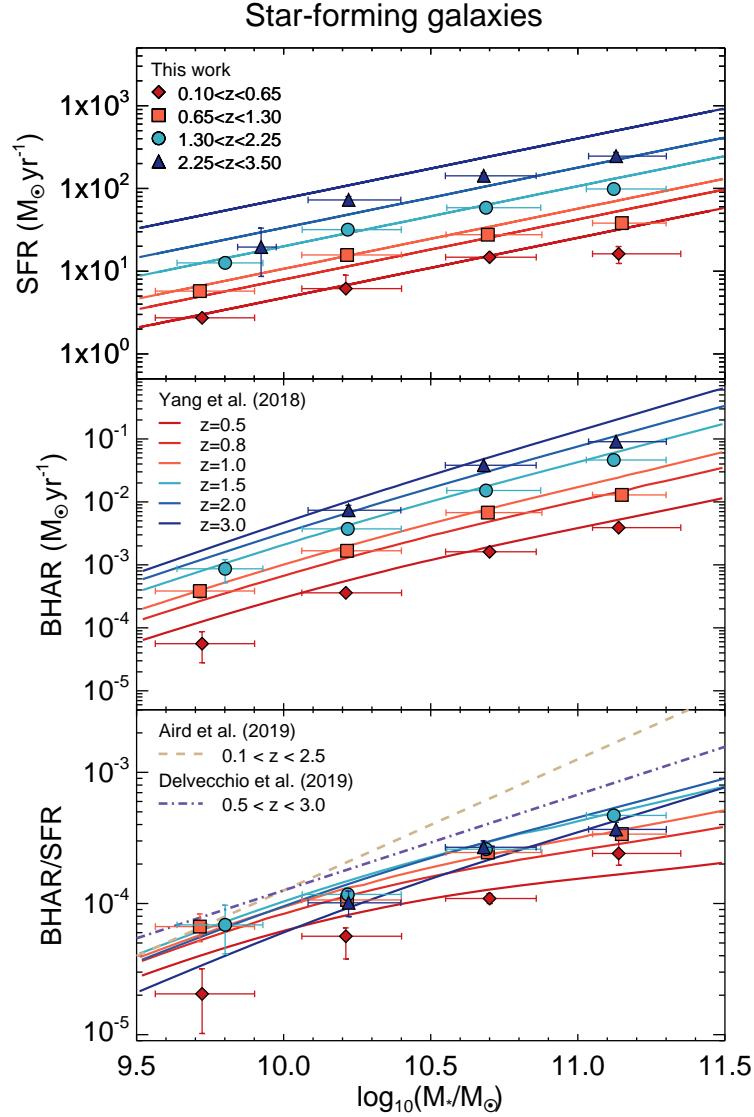


Figure 2.4: Comparison between our results for star-forming galaxies and data from Yang et al. (2018), Aird et al. (2019), and Delvecchio et al. (2019). Data points are taken from the left column of Figure 2.3, and the curves are adapted from Figure 14 in Yang et al. (2018) and were scaled to the same k -correction as we adopted here. Data from Aird et al. (2019) are taken from Fig. 13.

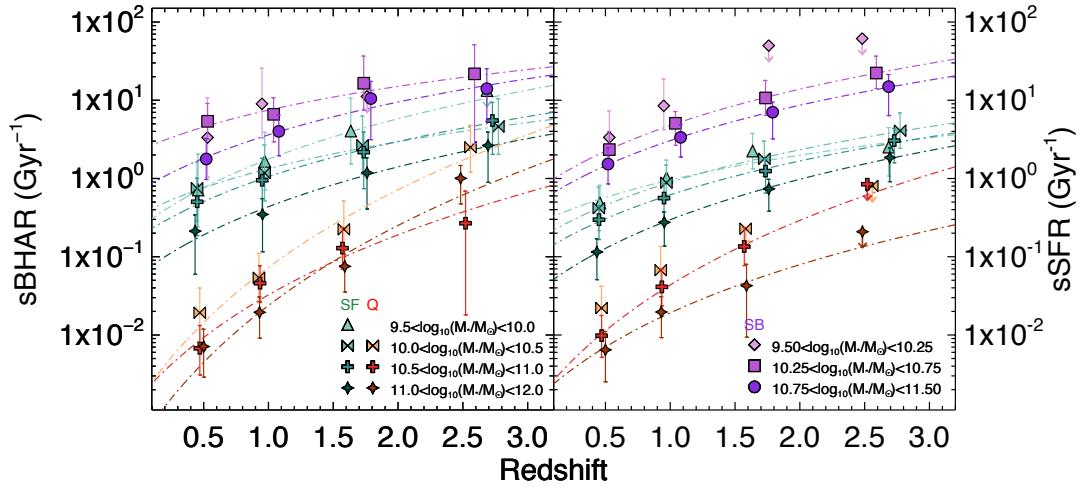


Figure 2.5: Specific black hole accretion rate (left panel) and sSFR (right panel) as a function of redshift for star-forming (SF, in green), quiescent (Q, in red), and starburst (SB, in purple) galaxies. The data points are placed at the median redshift of each mass bin and are coded in shape and color according to their median mass and type. Error bars represent the 90% confidence interval associated with each measure. We show an upper limit (the 95th percentile) for the data points whose sBHAR/sSFR 5th percentile is compatible with zero and for starburst galaxies at high redshift and low mass, where our sample is incomplete. The dot-dashed line is the best fit of the data to the curve $sBHAR(sSFR) = \delta(1+z)^\gamma$ and is color-coded according to the galaxy type and mass bin.

2.6.1 Black hole mass estimate

In order to estimate the black hole mass M_{BH} , we considered the fits in the bottom panels of Fig. 2.3 and in the right column of Table 2.3, obtained from the equation

$$\log \frac{\text{BHAR}}{\text{SFR}} = \alpha \log M_* + \beta. \quad (2.7)$$

The general equation we used for the computation is

$$\log \frac{\text{BHAR}}{\text{SFR}} = \log \frac{\dot{M}_{\text{BH}}}{M_*} = \log \frac{\partial M_{\text{BH}}}{\partial M_*} = \alpha \log M_* + \beta, \quad (2.8)$$

which is valid only under the assumption that the growth rates of stellar and black hole mass are constant in time. The ratios we measure show almost no evolution up to $z \sim 3.5$, which appears to justify this assumption. By integrating Equation 2.8 with respect to stellar mass, we obtained the estimate of the M_{BH} of each stellar mass and redshift bin:

$$M_{\text{BH}} = \frac{10^\beta}{1 + \alpha} M_*^{1+\alpha}, \quad (2.9)$$

where α and β are the same parameters in Table 2.3 and are the parameters we used in the computation. It follows from Eq. 2.9 that the black hole mass has a superlinear dependence on stellar mass in star-forming and starburst galaxies, but M_{BH} has a linear dependence on M_* in quiescent galaxies.

If instead of performing an indefinite integral we integrate between the initial masses ($M_{*,i}$ and $M_{\text{BH},i}$) and the final (observed) masses, the integrated Equation 2.8 would read

$$M_{\text{BH}} = \frac{10^\beta}{1 + \alpha} (M_*^{1+\alpha} - M_{*,i}^{1+\alpha}) + M_{\text{BH},i}. \quad (2.10)$$

We assumed that the initial black hole mass is lower by at least an order of magnitude ($\lesssim 10^5 M_*$) than the final mass, however, therefore it is negligible. The initial stellar mass instead eludes us, and because we subtracted it, our M_{BH} is an upper limit. We therefore decided not to use the complete expression of M_{BH} in Eq. 2.10, but the expression in Eq. 2.9.

2.6.2 Results for the specific accretions

Figure 2.5 shows the redshift evolution of the sBHAR (estimated using M_{BH} from Eq. 2.9) and sSFR for star-forming, quiescent, and starburst galaxies. sBHAR and sSFR have a decreasing trend toward lower redshift for the three galaxy types, and

the stellar mass shows a split: low-mass galaxies have higher values than high-mass galaxies, although this split is often not significant within the error bars.

These trends confirm what has been reported before about these specific accretions. They also provide new interesting insight. The spread in stellar mass is consistent with downsizing: it implies that high-mass galaxies have accreted most of their M_* and M_{BH} at high redshift and their accretion decreased fast and steeply, whereas low-mass galaxies have accreted their mass more slowly, but their accretion rate decreased more slowly with time. Downsizing has previously been observed for stellar mass (Cowie et al., 1996; Cimatti et al., 2006) and black hole luminosity (e.g., Marconi et al. 2004 but also Shankar et al. 2004, 2009; Aversa et al. 2015a, who reached similar conclusions from continuity equation arguments). In our plots, downsizing can be observed in the steepening of the slope γ at higher masses in star-forming and starburst galaxies and in the two specific accretions, which indicate a faster decrease for higher masses. Downsizing can also be seen in high-mass galaxies (darker data points), which show lower sSFR(sBHAR) on average than lower mass galaxies. This means that at even at higher redshifts, the most massive galaxies have already accreted most of their stellar and black hole mass, even though within the error bars, sSFR (sBHAR) data points are often compatible with a unique value for all stellar masses at a given redshift. In particular, this trend would not be present in star-forming galaxies without introducing the dependence of the black hole mass on the slope α from Equation 2.7.

We show in Figure 2.5 the best fits to the data when a redshift evolution of the form $s\text{BHAR} (\text{sSFR}) = \delta (1 - z)^\gamma$ is adopted, which we applied to each mass bin for all galaxy types when at least three data points were available³. The sSFR and sBHAR for star-forming and starburst galaxies are compatible with $\gamma = 2.8$, as in Sargent et al. (2012), within 1σ , but this is not the case for quiescent galaxies, which have higher slopes with a steeper decreases in specific accretions in time. Interestingly, we do not note a significant exponent difference between sBHAR and sSFR at given galaxy type. When we consider that there may be a small contribution from misclassified star-forming galaxies in the quiescent sample, especially at high redshift, the real trend may be even flatter, and approach a slope $\gamma = 2.8$. We do see a difference in the normalizations of the relations, which show lower specific accretions for quiescent galaxies, followed by star-forming galaxies, and finally, starburst galaxies, which tend to have higher specific accretions. When the normalizations for the two specific accretions are compared, they are very similar in the case of starburst galaxies,

³Fits performed with IDL/MPFIT (Markwardt, 2009)

while star-forming galaxies have a higher normalization for the sBHAR and quiescent galaxies show higher sBHAR than sSFR at high z .

Because the inverse of sSFR and sBHAR can be considered as the mass-doubling timescale of the M_* and of the SMBH, this means that the stellar mass of the galaxy and the SMBH are accreted faster in starburst and star-forming galaxies, which especially at high redshift are still efficient and can quickly double their mass. Quiescent galaxies are instead slower and have mass-doubling timescales of about a Hubble time or more. The black hole mass-doubling time of star-forming galaxies instead seems to be shorter than the stellar mass black hole mass-doubling time at every redshift and mass bin.

These similar evolution trends between sBHAR and sSFR strongly suggest a connection between the two accretions that appears to be present in all galaxy life phases. Star-forming galaxies appear to dominate the accretion histories; they are the most numerous galaxies and are able to substantially accrete their stellar and black hole masses. Starburst galaxies, with a higher capability of accretion but short-lived episodes, and quiescent galaxies, although their accretion capabilities are lower, appear to be able to accrete their black hole more efficiently than their stellar mass.

2.7 Relation between stellar mass and black hole mass

In Figure 2.6 we show our best-fit resulting $M_{\text{BH}}-M_*$ relations for star-forming galaxies derived from Eq. 2.9 with parameters from Table 2.3. These are superlinear relations with approximately $M_{\text{BH}} \propto M_*^{1.6}$. Our stellar masses are to be considered lower limits because as pointed out in Bell et al. (2003), stellar masses obtained using Bruzual & Charlot (2003) models lead to lower mass-to-light ratios and therefore to systematically lower stellar mass estimates at a given luminosity. A possible correction in our stellar masses that would cause them to agree better with those of Bell et al. (2003) (increase of ~ 0.25 dex) would further support our results.

We note that recent scaling relations proposed by other groups for both early-type galaxies and late-type galaxies show different scaling slopes and normalizations. Sahu et al. (2019) find a relation for early-type galaxies that is similar in both slope and normalization to the lines of Savorgnan & Graham (who used bulge stellar masses. See Shankar et al. (2016, 2019) for details) reported in Fig. 2.6. For late-type galaxies,

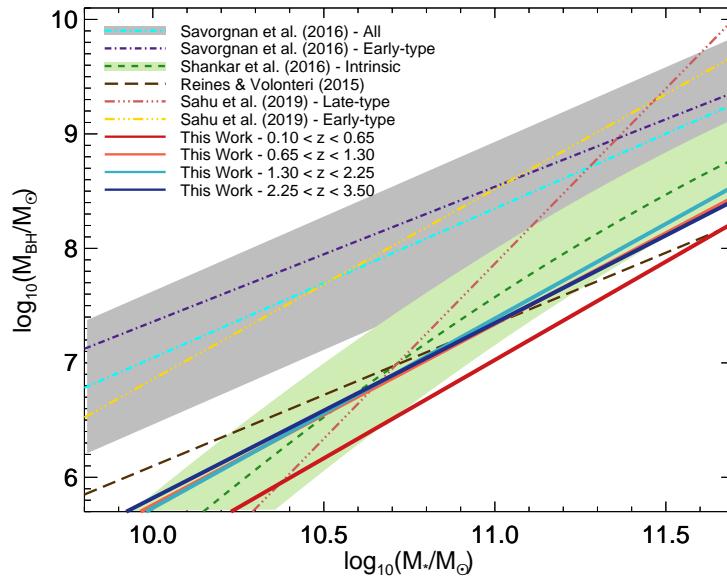


Figure 2.6: Comparison between our M_* and M_{BH} from Eq. 2.9 and parameters α and β from Table 2.3 for star-forming galaxies and results from models in the literature. Data from this work are shown as solid lines and are color-coded according to redshift as in Fig. 2.3 and 2.4. The relations from Savorgnan & Graham (2016) use the M_* of the bulge. The gray shaded area is the confidence range for Savorgnan & Graham (2016), and the light green are is the confidence range for Shankar et al. (2016).

Table 2.4: Fit parameters of the relations in Fig. 2.5. The equation we used is sBHAR (sSFR) = $\delta (1+z)^\gamma$.

Mass bins $\log_{10}(M_*/M_\odot)$	sBHAR parameters		sSFR parameters	
	δ	γ	δ	γ
Star-forming				
9.5–10.0	0.2 ± 0.3	2.9 ± 1.8	0.28 ± 0.13	1.8 ± 0.5
10.0–10.5	0.3 ± 0.3	2.0 ± 0.9	0.17 ± 0.09	2.4 ± 0.6
10.5–11.0	0.18 ± 0.14	2.6 ± 0.8	0.11 ± 0.05	2.5 ± 0.5
11.0–12.0	0.06 ± 0.04	2.9 ± 0.7	0.039 ± 0.016	2.9 ± 0.4
Quiescent				
10.0–10.5	0.0015 ± 0.0012	5.6 ± 0.9	-	-
10.5–11.0	0.0016 ± 0.0009	4.4 ± 0.7	0.0017 ± 0.0012	4.7 ± 1.0
11.0–12.0	0.0004 ± 0.0003	5.8 ± 0.8	0.0017 ± 0.0013	3.5 ± 1.1
Starburst				
9.50–10.25	-	-	-	-
10.25–10.75	2.3 ± 1.6	1.7 ± 0.9	0.8 ± 0.4	2.6 ± 0.5
10.75–11.50	0.7 ± 0.4	2.4 ± 0.6	0.5 ± 0.2	2.6 ± 0.5

Sahu et al. (2019) find a lower normalization, more consistent with our relation around $M_* \sim 10^{10.5} M_\odot$, but a much steeper slope of $\sim 3.0 \pm 0.5$, about 3σ steeper than ours. All in all our results point to either lower normalizations or flatter slopes than those identified from local dynamically measured SMBHs. On the other hand, our resulting scaling relations have a similar normalization with those proposed by Reines & Volonteri (2015) from local broad line AGN, despite them finding a close to linear relation, and agree well with the intrinsic scaling relation by Shankar et al. (2016). The latter have suggested that the local sample of (mainly) early-type galaxies with dynamically measured SMBH may be biased because a preselection might favor galaxies with the highest black hole masses and related gravitational radii. As recently suggested by Shankar et al. (2019), AGN samples that are clearly not affected by resolution-related selection effects should be closer to the intrinsic scaling relations. Our study further supports this view.

We note that our normalization is inversely proportional to the radiative efficiency assumed in Eq. 2.5. A match to the normalization of the local $M_{\text{BH}}\text{-}M_*$ relation by Savorgnan & Graham would require mean radiative efficiencies that are an order of

magnitude lower. This is disfavored on other grounds (Shankar et al., 2020b).

Our analysis adds key evidence, using IR data, to an increasing body of work (Delvecchio et al., 2019; Ding et al., 2020; Shankar et al., 2020b; Suh et al., 2020) on the weak evolution in the BHAR/SFR and its relatively low normalization relative to local raw dynamical $M_{\text{BH}}\text{-}M_*$ relations.

2.8 Conclusions

We performed a statistical study on the COSMOS field in order to constrain the history and coevolution of star formation and black hole accretion in star-forming, quiescent, and starburst galaxies. We selected a mass-complete sample from the COSMOS 2015 catalog (Laigle et al., 2016) by classifying normal star-forming and quiescent galaxies through the $\text{NUV}-r/r - J$ color-color diagram, and we selected starburst galaxies from the *Herschel*-selected sample in Gruppioni et al. (2013). We performed an X-ray stacking analysis and combined it with detected sources from Civano et al. (2016) in order to estimate the average X-ray luminosity and therefore average BHAR. We estimated the SFR from FIR stacking and UV SED fitting. Our main results are listed below.

1. We find a robust $L_X - M_*$ relation for star-forming galaxies at all considered redshifts. This relation evolves with an increasing normalization at higher redshifts. The X-ray luminosity of quiescent galaxies is close to that of star-forming galaxies, especially at low masses. At high masses, L_X in quiescent galaxies is lower than in star-forming galaxies. The X-ray luminosities of starburst galaxies are compatible with star-forming at high redshifts and evolve mildly down to low redshift, where they are clearly higher than those of star-forming galaxies.
2. The $L_X - M_*$ relation translates into a BHAR- M_* relation in Fig. 2.3 (middle row) that shows that the evolution of the BHAR in star-forming galaxies is faster at lower redshifts, it has a pronounced redshift evolution, and a weak mass dependence in quiescent galaxies. In turn, starburst galaxies have a marked mass dependence and a distinctive redshift evolution: going back in time, it reaches a maximum at $z \sim 1.7$ to then decrease again.
3. BHAR in star-forming galaxies increases more with stellar mass than the SFR. In quiescent galaxies, the BHAR values lie close to the BHAR of star-forming

galaxies, while the SFR of quiescent galaxies is clearly below the main-sequence. It is interesting that while the SFR of starburst galaxies continues to increase at higher redshifts, at the highest redshift in our study, the BHAR of these galaxies decreases.

4. The ratio between BHAR and SFR in star-forming and starburst galaxies has a positive relation M_* that is almost time-independent. The ratio is higher for quiescent galaxies, compatible with a flat trend in M_* , indicating a stronger tendency for this type of galaxy to accrete onto the black hole than to form stars, regardless of stellar mass. From this it follows that M_{BH} has a superlinear dependence on M_* in star-forming and starburst galaxies, and the dependence is linear in quiescent galaxies.
5. sBHAR and sSFR follow very similar decreasing trends in time. We see signs of downsizing in all types of galaxies, a faster accretion (of M_{BH} and M_*) in starbursts followed by star-forming galaxies, and finally, by quiescent galaxies with mass-doubling timescales of about a Hubble time.
6. The resulting $M_{\text{BH}}-M_*$ relation from our data agrees well with independent determinations of the relation that were retrieved from AGN samples and Monte Carlo simulations.

All of these results confirm the coevolution of host galaxy and black hole follows the pattern of downsizing at all redshifts and in different galaxy evolutionary phases. In this picture, the bulk of the black hole and stellar masses is accreted in galaxies during the main-sequence phase through secular processes, where more massive galaxies are more efficient at accreting the black hole. Starburst episodes play a lesser role for both accretions because only a few galaxies are in this phase and these episodes are only weakly able to enhance black hole accretion at high redshifts. The deeper potential well of more massive and possibly more compact galaxies seems to be playing a role in feeding the black hole more efficiently in star-forming and starburst galaxies. In the quiescent life phase of galaxies, the black hole accretion is not as penalized as the star formation. The gas availability reported by Gobat et al. (2018) means that this gas may not go to star formation because of different galactic properties in the different life phases (e.g., disk and bulge dynamics), but to accrete the black hole. Finally, we find additional evidence that suggests that the $M_{\text{BH}}-M_*$ may have a lower normalization than the local dynamical relation.

CHAPTER 3

A semi-empirical model point of view

The work in this chapter has been submitted for publication on the main journal of the Monthly Notices of the Royal Astronomical Society.

In this Chapter we aim to expand the results from Chapter 2 and understand how they originate by taking advantage of semi-empirical models (SEMs) and the sample selected in there.

Semi-empirical models are a competitive, fast and flexible methodology, extensively used in recent years to constrain the degree of evolution and mergers in galaxies, as well as the degree of coevolution with their central BHs (Conroy & White, 2013; Georgakakis et al., 2019; Comparat et al., 2019; Aird & Coil, 2021; Shankar et al., 2020a; Allevato et al., 2021). It is particularly relevant the application of SEMs to the creation of active and normal galaxy “mock” catalogs (e.g. Georgakakis et al., 2019; Comparat et al., 2019; Aird & Coil, 2021; Shankar et al., 2020a; Allevato et al., 2021), which are a vital component of the planning of imminent extra-galactic surveys such as Euclid (Laureijs et al., 2011). The advantage of this methodology is that by using few input relations, it creates a mock catalog of active and normal galaxies that – by design –

reproduces the X-ray luminosity function of active galaxies. Moreover, these models allow us to disentangle the effects on relevant relations of the input parameters, shedding light on the processes controlling the co-evolution of BHs and their hosts.

In this Chapter we perform new estimates for the X-ray detected sample selected in Chapter 2 and use comprehensive semi-empirical mock catalogs of active BHs to pin down which parameters control the shape and evolution of the $\langle L_X \rangle - M_*$ relation. We explore a variety of inputs in our model, such as the shape of the Eddington ratio distribution $P(\log \lambda)$ which carries information on the accretion of a BH, or the normalization of the $M_{\text{BH}} - M_*$ scaling relation. More specifically, we find that its slope and normalization are mostly determined by, respectively, the $M_{\text{BH}} - M_*$ relation and mean Eddington ratio.

In Section 3.1 we present our model and in Section 3.2 we highlight the main parameters controlling the $L_X - M_*$ evolution at different redshifts and galaxy phases. In Sections 3.3, and 3.4 we discuss our findings and draw our conclusions on their relevance to the galaxy-BH co-evolution.

3.1 Building robust AGN mock catalogs

In this study we create realistic mock catalogs of AGN and non-active galaxies to study which input parameters mostly control the $\langle L_X \rangle - M_*$ relation at different redshifts. Below provide the most relevant steps in the generation of our mocks, and refer the reader to Allevato et al. (2021) where the model is presented, for full details. In the work in this Chapter we run the model testing and tuning the entire range of input parameters. We also perform the analysis of the model output, the galaxy mock catalogs, in order to be able to compare the mocks to our observables.

The first step for the creation of mocks consists in generating a halo distribution via a halo mass function from Tinker et al. (2008) at the redshift of interest. To each dark matter halo we assign a galaxy stellar mass via abundance matching techniques¹, using the relation of Moster et al. (2010) with updated parameters from Grylls et al. (2019, Eq. 5) with a normal scatter in stellar mass at fixed halo mass of 0.11 dex. We then assign a BH mass via the empirically calibrated $M_{\text{BH}} - M_*$ relation by Reines &

¹We immediately note that the exact choices for this first step of the mock generation are irrelevant to our results and conclusions discussed below. As in this work we are not studying the environment of active galaxies, the information on host halo mass is here only given for completeness. Equivalent mocks could be generated by simply extracting galaxies from an input stellar mass function.

Volonteri (2015), with an intrinsic scatter of 0.55 dex, and also explore the impact of adopting other $M_{\text{BH}} - M_*$ relations from Shankar et al. (2016), Davis et al. (2018) and Sahu et al. (2019), which bracket the systematic uncertainties in the BH-galaxy stellar mass in the local Universe. We then assume that each relation does not evolve with redshift, as suggested by a number of studies (e.g. Shankar et al., 2009; Delvecchio et al., 2019; Suh et al., 2020; Shankar et al., 2020b, and Fig. 2.6). To each galaxy and BH we then assign an Eddington ratio $\lambda \equiv L_{\text{bol}} / L_{\text{Edd}}$ and convert bolometric luminosities L_{bol} to intrinsic (i.e., unobscured) 2-10 keV X-ray luminosities L_X via the same bolometric corrections k_X adopted in Chapter 2 (see Sec. 2.3.2). Following the formalism in, e.g., Shankar et al. (2013) and Allevato et al. (2021) and references therein, which in turn follows the one routinely adopted in continuity equation models, the AGN luminosity function at any given redshift z can be expressed by the convolution

$$\Phi(\log L_{\text{bol}}, z) = \int_{\log \lambda_{\min}}^{\log \lambda_{\max}} U(y, z) n(y, z) P(\log \lambda, z) d \log \lambda \quad (3.1)$$

where $y = \log M_{\text{BH}}$ and $n(y, z)$ is the total BH mass function. $P(\log \lambda, z)$ is the Eddington ratio distribution, which we assume for simplicity to be independent of BH mass, normalized to unity in the range $\log \lambda_{\min} < \log \lambda < \log \lambda_{\max}$. $U(y, z)$ is the *intrinsic* duty cycle, i.e., the fraction of all black holes of mass y that are active and accreting mass at an Eddington rate in the range $\log \lambda_{\min} < \log \lambda < \log \lambda_{\max}$ at redshift z . We set our minimum Eddington ratio to $\log \lambda_{\min} = -4$ and the maximum Eddington ratio to $\log \lambda_{\max} = 1$, noticing that the exact value chosen for $\log \lambda_{\max}$ does not alter any of our results as the adopted Eddington ratio distributions have extremely low probabilities above the Eddington limit.

The flexibility offered by Eq. 3.1 allows to disentangle the effects of the shape of $P(\log \lambda, z)$, which carries information on the accretion properties of a BH, from the fraction $U(y, z)$ of active BHs accreting above a certain threshold in Eddington ratio. The reference $P(\log \lambda, z)$ distribution is taken to be a simple Gaussian in $\log \lambda$ characterized by a standard deviation σ and a mean μ . We will show that the shape of the $P(\log \lambda, z)$ distribution plays a minor role in the outputs as long as the characteristic Eddington ratio, defined as

$$\zeta_c(z) \equiv \langle \log \lambda \rangle(z) = \int_{\log \lambda_{\min}}^{\log \lambda_{\max}} P(\log \lambda, z) \log(\lambda) d \log(\lambda), \quad (3.2)$$

is the same. In Eq. 3.2 $\log \lambda_{L_X, \min}$ is the Eddington ratio corresponding to the limiting luminosity at that redshift. We assume a constant duty cycle of $U = 0.2$ as suggested by Goulding et al. (2010) from local X-ray AGN, but we also explore the impact on our

results of varying the input duty cycle with BH mass, specifically decreasing with M_{BH} as inferred by Schulze & Wisotzki (2010) and Schulze et al. (2015), and also increasing with M_{BH} , as proposed by Man et al. (2019). Although these works were based on AGN samples with different selections, we use these duty cycles simply as a guidance to explore the impact on our results of different “shapes” of the input intrinsic duty cycles $U(y, z)$.

When comparing with the data we must retain from the full BH mock only those active BHs shining above the X-ray flux limit of the observational survey (e.g., Shankar et al., 2013). In our reference sample, the *Chandra* COSMOS Legacy Survey (COSMOS-Legacy, Civano et al., 2016), the X-ray flux limit corresponds to luminosities of $L_X = 10^{42}$ erg/s in the lower redshift bin ($z = 0.45$), increasing by an order of magnitude or more at higher redshifts (see below for details). When computing all AGN-related observational probes, such as the AGN luminosity function (Eq. 3.1), the characteristic Eddington ratio $\zeta_c(z)$ (Eq. 3.2), or the mean X-ray luminosity (Eq. 3.5), we thus include only those active black holes shining above the *Chandra* COSMOS Legacy Survey flux limit at the given redshift². For example, although we fix our minimum Eddington ratio to $\log \lambda_{\min} = -4$ for our input $P(\log \lambda, z)$ (e.g., in Eq. 3.1), after imposing the cut in X-ray flux limit, among the BHs with mass $M_{\text{BH}} \lesssim 10^8 M_\odot$ in the lowest redshift bin, only those accreting at an Eddington rate $\log \lambda_{X,\min} \gtrsim -3$ will be included in the comparison with the data. We will discuss below that the flux limit plays a non-negligible role when comparing theoretical AGN mocks to observations, particularly with respect to the observed fraction of active black holes as a function of host galaxy stellar mass (Figure 3.3).

We assign SFRs to quiescent, normal star-forming, and starburst galaxies based on their respective SFR- M_* relation. For starburst and quiescent galaxies, we adopt the SFR fits from Table 2.3, while for the “main sequence” we adopt the Schreiber et al. (2015, Eq. 9) flexible parametric formula

$$\log_{10} \left(\frac{\text{SFR}}{M_\odot \text{yr}^{-1}} \right) = m - m_0 + a_0 r - a_1 [\max(0, m - m_1 - a_2 r)]^2 \quad (3.3)$$

with $m \equiv \log_{10}(M_*) - 9$ and $r \equiv \log_{10}(z + 1)$. Best-fit parameters for our COSMOS

²The COSMOS field is a great combination of area and depth making it ideally suited to probe the accretion properties of active BHs. A deeper field may be more sensitive to the faint end shape of the Eddington ratio distribution, but would not allow to include more luminous sources. A shallower field on the other hand, may return better statistics for the more luminous sources, but rapidly losing the fainter ones.

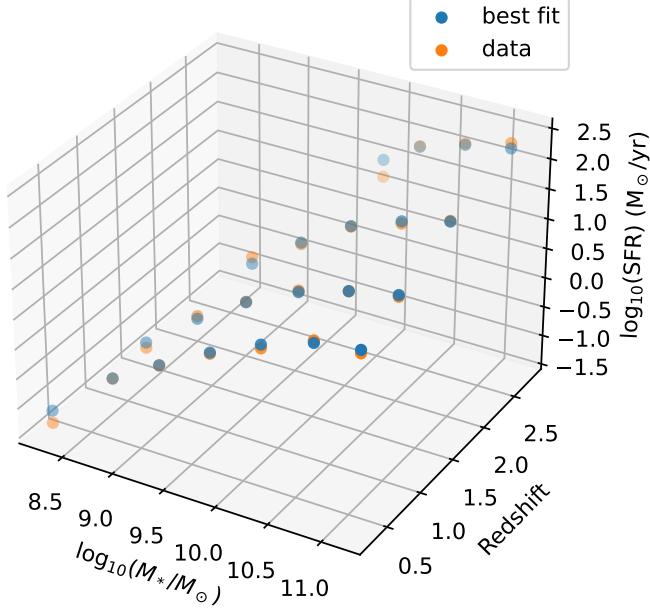


Figure 3.1: A 2-dimensional plot of the SFR data for star forming galaxies from Chapter 2 as a function of M_* and redshift, compared with our best fit of Eq. 3.3.

data are $a_0 = 2.29 \pm 0.12$, $a_1 = 0.25 \pm 0.04$, $a_2 = 0.33 \pm 0.30$, $m_0 = 0.64 \pm 0.03$, $m_1 = 0.55 \pm 0.11$, and the fit is shown in Fig. 3.1. We add a dispersion of 0.2 dex to the SFR.

Irrespective of their duty cycle, we assign to each galaxy in the mock an X-ray luminosity from X-ray binary emission following Lehmer et al. (2016, Table 3), and when computing the average X-ray luminosity competing to a given bin of stellar mass, we then subtract the mean binary emission competing to that bin of stellar mass and star formation rate as in Sec. 2.3.2. We note that neglecting X-ray binary emission entirely from our procedure would yield very similar results. Following the procedure described above, we generate diverse galaxy mock catalogs with distinct choices of the input $M_{\text{BH}} - M_*$ scaling relations, duty cycles, and $P(\log \lambda, z)$ distributions. We then divide each AGN mock catalog in bins of stellar mass, and select the BHs that shine above the flux limit of the COSMOS-Legacy survey (Marchesi et al., 2016), i.e., the “detected” sources of the mock, as discussed above. The first observable we compute is the AGN fraction, defined as

$$\text{AGN fraction}(M_*, z) = \frac{\sum_i U_{i,\text{detected}}}{N_{\text{tot}}}, \quad (3.4)$$

where the sum in the numerator runs over all active BHs above the flux limit, and N_{tot} at the denominator is the total number of active and normal galaxies in the specified

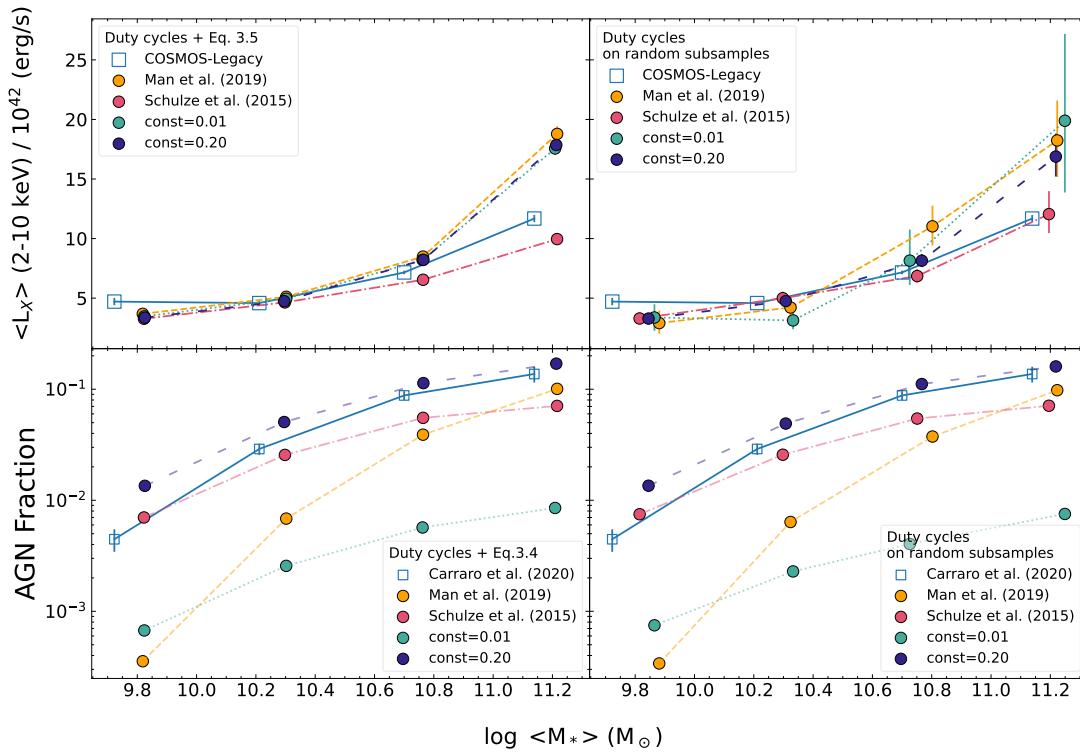


Figure 3.2: Comparison of $\langle L_X \rangle$ and AGN fraction obtained with various input duty cycles as in our approach in the rest of the Chapter (left panels, using Equations 3.4, and 3.5) and by randomly assigning a star formation category and AGN activity flag.

stellar mass bin. We note that the probability for a galaxy to be detected above a certain X-ray luminosity threshold, i.e., the “observed” duty cycle, will depend not only on the assumed (intrinsic) duty cycle, but also on other properties such as its BH mass and Eddington ratio. We will discuss below the differences between observed and intrinsic duty cycles and highlight how different input parameters in the mocks can generate similar observed fractions of AGN. The comparison between the observed AGN fraction and predicted input duty cycle $U(\log M_{\text{BH}}, z)$ yields important constraints on the accretion properties of active BHs when coupled to other observables, as we will discuss below. Finally, we perform 500 bootstraps out of which we extract the median SFR and M_* , and the linear mean L_X weighted by the AGN duty cycle

$$\langle L_X \rangle = \frac{\sum_i U_i(y_i, z) L_X(y_i)}{\sum_i U_i(y_i, z)}, \quad (3.5)$$

where $\log L_X(y_i) = 38.1 + \log \lambda_i + y_i - \log k_X$, where again the sums run over all detected BHs in the selected stellar mass bin. The key advantage of computing mean X-ray luminosities only considering sources above the flux limit, is that it provides a tracer of BH luminosity largely independent of the duty cycle, as demonstrated below. For each bootstrapped distribution we compute the median SFR and $\langle L_X \rangle$ with their 5th and 95th percentiles, following the same procedure as in the comparison observational sample selected in Chapter 2. We note that in our work in the aforementioned Chapter, the mean X-ray luminosities were computed over the full sample of M_* -selected galaxies, including both detected sources and stacking on non-detected sources. Eq. 3.5 is instead a weighted mean over only the detected sources, and thus we recomputed the mean X-ray luminosities in Chapter 2 sample limiting the analysis to only X-ray detected sources. While in Chapter 2 we assumed an average characteristic obscuration/extinction correction for all sources competing to a given bin of X-ray luminosity, we here apply to each individual source the obscuration correction listed in the Marchesi et al. (2016) catalog.

3.2 Results

3.2.1 Reproducing the measured fraction of detected galaxies

Before showing the results on the predicted mean X-ray luminosity of detected galaxies, we discuss if and when our model is able to match the fraction of X-ray sources directly observed in COSMOS-Legacy as a function of stellar mass. The open blue

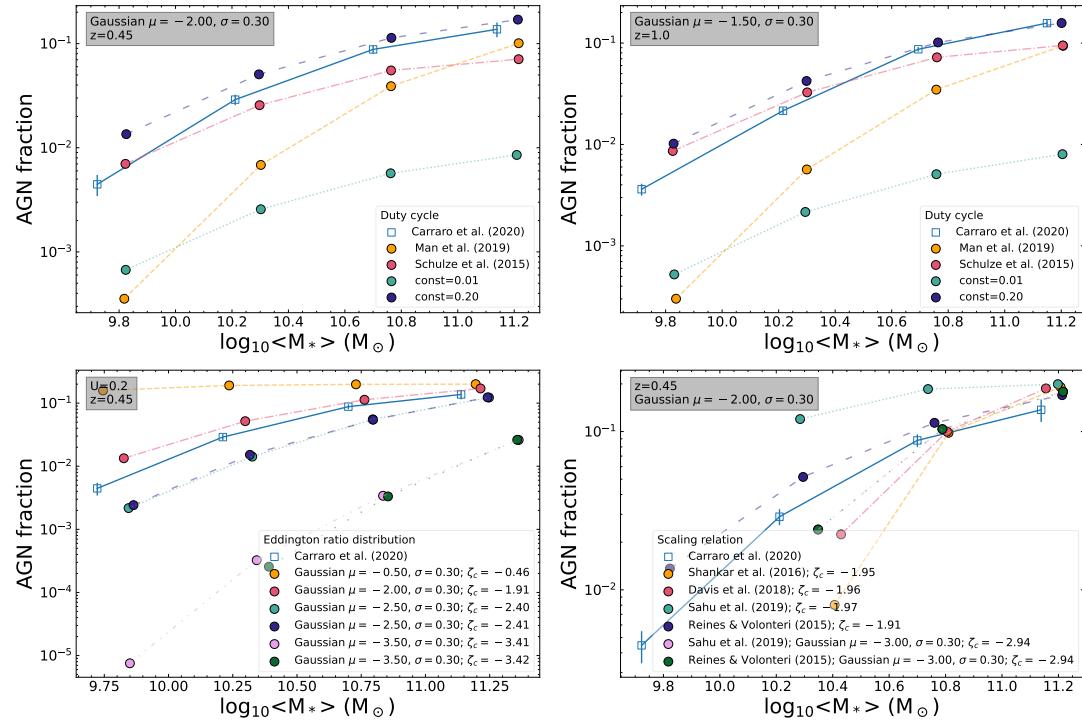


Figure 3.3: Dependence of the fraction of X-ray detected galaxies (AGN fraction) on the input model duty cycle (top panels), Eddington ratio distribution (bottom-left panel) and scaling relation (bottom-right panel)

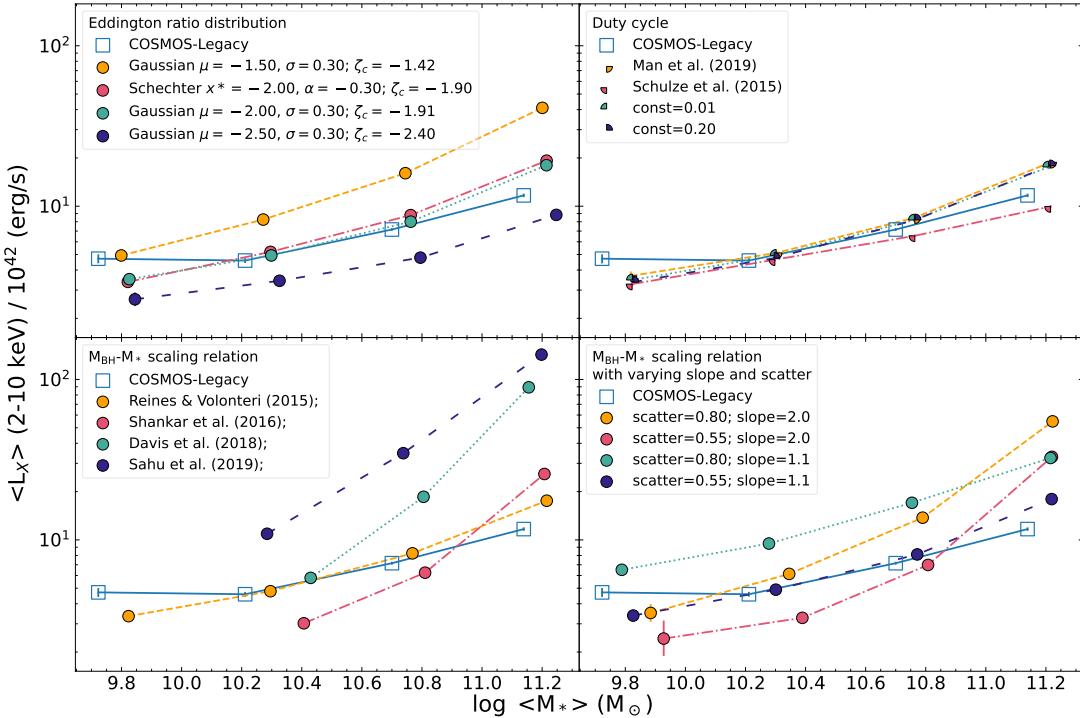


Figure 3.4: A gallery of detected $L_X - M_*$ relations of detected sources at $z = 0.45$ for star forming galaxies obtained by varying one of the input relations at a time. The relation that varies in each subplot is reported in the legend. Results from COSMOS X-ray detected sources at the same redshift from Chapter 2 are included in all plots for comparison (blue squares). Top left: $L_X - M_*$ relation obtained by changing the Eddington ratio distribution function. We use a Schechter function and Gaussian function in $\log(\lambda)$ with varying mean μ and standard deviation σ values. Top right: $L_X - M_*$ relation obtained by changing the duty cycle method. Bottom left: $L_X - M_*$ relation obtained by changing the $M_{\text{BH}} - M_*$ scaling relation. Each scaling relation is shown within its original stellar mass range of derivation. Bottom right: $L_X - M_*$ relation obtained with a toy $M_{\text{BH}} - M_*$ scaling relation where we change the logarithmic slope β of the relation $\log M_{\text{BH}} = \alpha + \beta \log M_*$ and increase its scatter. Original Reines & Volonteri (2015) values are: $\beta = 1.1$ and 0.55 dex scatter.

squares in Figure 3.3 are the COSMOS-Legacy data, taken from Tables A.1. to which we associate a binomial error on the number of detected AGN and a Poisson error on the total number of sources, combined together with standard error propagation applied to $\frac{N_{\text{det}}}{N_{\text{tot}}}$.

We then compare the data with our models filtered by the flux limit of the observations, which is equal to $L_{X,\min} = 10^{42} \text{erg/s}$ and $L_{X,\min} = 6 \times 10^{42} \text{erg/s}$ at $z = 0.45$ and $z = 1.0$, respectively. We adopt as our reference model one characterized by a constant input duty cycle of $U = 0.2$, a Gaussian Eddington ratio distribution in $\log \lambda$ peaked at $\mu = -2$, and the $M_{\text{BH}} - M_*$ scaling relation from Reines & Volonteri (2015). We will show below that this choice of input parameters provides a good match to both the mean AGN X-ray luminosity and AGN luminosity function. We then vary several of the input parameters, starting from the duty cycle at both $z = 0.45$ and $z = 1$ (left and right top panels, respectively), the peak of the Gaussian $P(\log \lambda, z)$ (bottom, left panel), and the input $M_{\text{BH}} - M_*$ scaling relation (bottom, right panel). It is first of all interesting to note from the top panels that, once the Gaussian $P(\log \lambda, z)$ and $M_{\text{BH}} - M_*$ scaling relation are fixed to our reference choices, the data are consistent with an input duty cycle $U \sim 0.2$ constant in both stellar/black hole mass and redshift, at least up to $z \lesssim 1$ (dark blue dashed lines in both top panels). The apparent strong increase of the AGN fraction with stellar mass is simply induced by the imposed flux limit. A too strong mass dependence in the input duty cycle, as suggested by the local fraction of optical AGN measured by Man et al. (2019) in SDSS, would be inconsistent with the data (dashed, orange lines), as well as an overall too low initial fraction (dotted, turquoise lines with $U = 0.01$).

The bottom left panel of Figure 3.3 shows that a varying input $P(\log \lambda, z)$ distribution, and thus a varying characteristic ζ_c , as labeled, generates widely different AGN fractions. More specifically, the higher the ζ_c the more luminous are, on average, the mock AGN, which in turn implies that proportionally less sources are removed by the cut imposed by the flux limit. We find that when $\zeta_c \gtrsim -0.5$, the observed AGN fraction is nearly identical to the input $U \sim 0.2$ (dashed, yellow line), while it rapidly diverges from the input $U \sim 0.2$ dropping towards lower mass, less luminous AGN when $\zeta_c \lesssim -2$. The right lower panel of Figure 3.3 also shows that a flatter or steeper $M_{\text{BH}} - M_*$ input scaling relation, such as the ones from dynamically measured M_{BH} by Sahu et al. (2019, dotted, turquoise line) in early type galaxies and Davis et al. (2018, dot-dashed, magenta line) in late type galaxies, naturally induce a proportionally flatter or steeper AGN fraction, because they map galaxies of same stellar mass

to more massive/more luminous or less massive/less luminous AGN. In conclusion, the observed AGN fraction can contribute to efficiently break the degeneracies in the input parameters (see also Section 3.3), and, when combined with other independent constraints on, e.g., the BH-galaxy scaling relations and/or the Eddington ratio distributions, it is a powerful diagnostic of the intrinsic AGN duty cycle $U(y, z)$, and it can thus be used to constrain the accretion history of supermassive black holes.

3.2.2 The effect of the model's inputs on the $\langle L_X \rangle - M_*$ relation

In Figure 3.4 we compare the mean X-ray luminosity of detected active galaxies in a given bin of stellar mass, which in what follows we will continue labeling simply as $\langle L_X \rangle$ (Eq. 3.5), with several different model predictions. To pin down the input parameters that mostly control the $\langle L_X \rangle - M_*$ relation, we explore in Figure 3.4 how the relation varies by changing, from top left to bottom right, the $P(\log \lambda, z)$, the duty cycle, the full $M_{\text{BH}} - M_*$ relation, and only the slope/scatter of the Reines & Volonteri (2015) relation, as labeled. All the mocks are generated at $z = 0.45$, though the results are applicable to all redshifts, as further discussed below. In Figure 3.4 the data refer to only the subsample of star forming, main sequence galaxies. As anticipated in Sec 3.1 and Eq. 3.5, the mean $\langle L_X \rangle$ should in principle be weighted by the fractional number of detected sources within a given star formation class (e.g., quiescent, star forming, starbursts). However, this additional weighting can be neglected as it is canceled out in Eq. 3.5, being a constant in each bin of stellar mass.

The top panels of Figure 3.4 clearly show that while the normalization of the $\langle L_X \rangle - M_*$ relation is strongly controlled by the characteristic Eddington ratio ζ_c (left panel), it has a negligible dependence on the input AGN duty cycle (right panel). This behavior is expected as the $\langle L_X \rangle$ in Eq. 3.5 is an average luminosity calculated only on the fraction of active sources, and as such it is largely independent of the number of BHs active in a given bin of stellar mass, but strongly dependent on the *rate* at which these BHs are accreting. We show in the top left panel of Figure 3.4 that a Schechter or Gaussian $P(\log \lambda, z)$ yield the same mean X-ray luminosity $\langle L_X \rangle$ at fixed stellar mass as long as their ζ_c are the same (dotted turquoise and dot-dashed magenta lines). It is indeed the characteristic Eddington ratio ζ_c , and not the overall shape of the $P(\log \lambda, z)$ input distribution, to determine the level of mean X-ray luminosity in active galaxies at fixed stellar mass and fixed $M_{\text{BH}} - M_*$ relation. Nevertheless, some constraints

even on the shape of the $P(\log \lambda, z)$ could be derived from our methodology. For example, assuming a steeper/flatter faint end in the input Schechter $P(\log \lambda, z)$ function, would induce a lower/higher ζ_c . To then preserve the same ζ_c necessary to match the observed $\langle L_X \rangle - M_*$ relation, would in turn require a shift in the knee of the Schechter function, and the new combination of faint end slope and knee can then be tested against the AGN luminosity function (which we further discuss below). It is relevant to reiterate at this point that the observations are only sensitive to Eddington ratios corresponding to luminosities above the survey flux limit (Eq. 3.2), and thus are sensitive only to portion of the $P(\log \lambda, z)$ above the minimum Eddington ratio detectable in the sample.

Some residual, weak dependence on the duty cycle may be visible in the right panel of Figure 3.4 especially towards higher stellar masses (dote-dashed, magenta line). This (tiny) dependence of the $\langle L_X \rangle - M_*$ relation on the input duty cycle is a simple byproduct of the scatter in the $M_{\text{BH}} - M_*$ relation and of our definition of input duty cycle: $U(\log M_{\text{BH}}, z)$ is dependent on BH mass, and thus at fixed stellar mass, a variety of BHs with different weights could contribute to the mean $\langle L_X \rangle$ slightly altering its final value depending on the shape (not the normalization) of the input duty cycle $U(\log M_{\text{BH}}, z)$.

The bottom left panel of Figure 3.4 shows instead a close link between the normalization of the input $M_{\text{BH}} - M_*$ relation and the normalization in the $\langle L_X \rangle - M_*$ relation: at fixed ζ_c , a lower $M_{\text{BH}} - M_*$ relation will result in a proportionally lower $\langle L_X \rangle - M_*$ relation, and vice versa. This causal link between the two relations naturally arises from the proportionality between X-ray luminosity and BH mass, which in turn is linked to the host galaxy stellar mass via the $M_{\text{BH}} - M_*$ relation. The right panel of Figure 3.4 shows the variations in the $\langle L_X \rangle - M_*$ relation for the same input $M_{\text{BH}} - M_*$ relation with varying slope or scatter, as labeled. A steeper/shallower $M_{\text{BH}} - M_*$ scaling relation will result in a proportionally steeper/shallower $\langle L_X \rangle - M_*$ relation, while a lower/higher scatter will decrease/increase the normalization of the $\langle L_X \rangle - M_*$ relation, mainly due to the lower/larger contribution of active BHs, especially the more massive and luminous ones. It is thus clear from Figure 3.4 that the slope and normalization of the input $M_{\text{BH}} - M_*$ relation, as well as the input ζ_c , all play a significant, and in fact degenerate, role in shaping the $\langle L_X \rangle - M_*$ relation. For example, a flatter slope in the $M_{\text{BH}} - M_*$ relation or a mass-dependent ζ_c , progressively decreasing at larger masses, could both produce a flatter slope in the $\langle L_X \rangle - M_*$ relation. Also, decreasing ζ_c with increasing BH mass could indeed reconcile the our

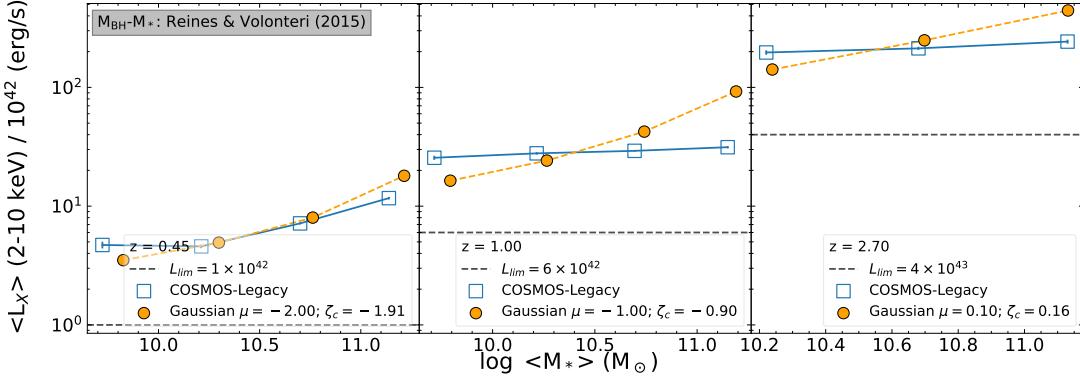


Figure 3.5: The $\langle L_X \rangle - M_*$ relations at $z = 0.45$ (left panels), $z = 1.0$ (central panels) and $z = 2.7$ (right panels) obtained by assuming a $M_{BH} - M_*$ scaling relation from Reines & Volonteri (2015) and a Gaussian in $\log(\lambda)$ with standard deviation $\sigma = 0.3$ dex. We vary the Eddington ratio distribution in order to reproduce the observational results from the COSMOS-Legacy detected sources selected in Chapter 2. Black dashed lines represent the survey luminosity limits.

observational results from Chapter 2 with a steeper $M_{BH} - M_*$ relation as calibrated in the local Universe (e.g., Shankar et al., 2016; Davis et al., 2018). If the scaling relation between BHs and their hosts is constrained via independent methods, such as AGN clustering (e.g., Shankar et al., 2020a; Allevato et al., 2021), then the $\langle L_X \rangle - M_*$ relation can be used to constrain the mean ζ_c as a function of galaxy stellar mass and redshift, as further discussed below.

3.2.3 Reproducing the $\langle L_X \rangle - M_*$ relation through cosmic time

In this Section we extend the comparison to data on the $\langle L_X \rangle - M_*$ relation at different redshifts. We showed in the previous Section that the $\langle L_X \rangle$ can provide valuable constraints on the mean Eddington ratio of active BHs. Thus, by studying the $\langle L_X \rangle - M_*$ at different redshifts and galaxy stellar masses, we can build a more comprehensive view of how BHs accrete at different epochs and in different host galaxies. The data point to a steady decrease of the mean $\langle L_X \rangle - M_*$ luminosity with cosmic time at fixed host galaxy stellar mass. As discussed above, this decreasing trend could be interpreted either as a progressive decline in the normalization of the $M_{BH} - M_*$ relation and/or in the characteristic ζ_c . The latest data suggest a rather weak evolution in the $M_{BH} - M_*$ relation up to at least $z \sim 2.5$ (e.g., Suh et al., 2020; Shankar et al., 2020b)

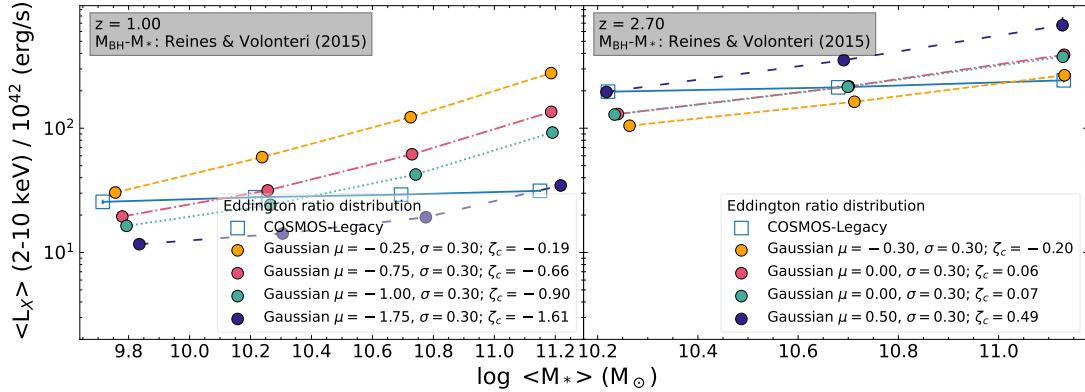


Figure 3.6: $L_X - M_*$ relation for active galaxies at $z = 1.0, 2.7$, looking for a ζ_c that fits each M_* bin.

thus favoring, in our approach, a steady decrease in ζ_c , which would also be in line with independent observations (Kollmeier et al., 2006) and continuity equation models (Shankar et al., 2013; Aversa et al., 2015b).

In Figure 3.5 we show the predicted $\langle L_X \rangle - M_*$ relation for mock catalogs at $z = 0.45, 1.0, 2.7$ (left, central and right panels respectively), generated by assuming as a reference the Reines & Volonteri (2015) $M_{\text{BH}} - M_*$ relation. At each redshift we plot the models with an input $P(\log \lambda, z)$ Gaussian distribution with a μ value (the corresponding ζ_c values are very similar being Gaussian distributions) chosen in a way to match the central value of the $\langle L_X \rangle - M_*$ distribution at each redshift. We find that, as can be better seen in Fig. 3.6, assuming a strictly constant $M_{\text{BH}} - M_*$ relation, to reproduce the data we would need a drop of a factor of $\gtrsim 100$ in the characteristic Eddington ratio ζ_c from $z \sim 2.7$ to $z \sim 0.45$, which mirrors the fast drop in mean Eddington ratio also derived in some observational data and continuity equation results (see, e.g., Fig. 12 in Shankar et al., 2013). We also note that at $z \gtrsim 1$, on the assumption that the input $M_{\text{BH}} - M_*$ relation remains constant in both slope and normalization, the models tend to produce a $\langle L_X \rangle - M_*$ relation steeper than what observed, which in turn would require a ζ_c decreasing with increasing stellar mass by a factor $\lesssim 3$ to improve the match to the data. A systematically lower mean Eddington ratio for more massive galaxies would imply that their more massive BHs should have grown faster, the so-called *downsizing* trend, already introduced and discussed in the previous Chapters, in which more massive galaxies/BHs form faster than less massive galaxies/BHs. The results in Figure 3.5 taken at face value suggest that BHs would be accreting close to their Eddington limit at $z \gtrsim 2.5$, and then rapidly shut off at lower

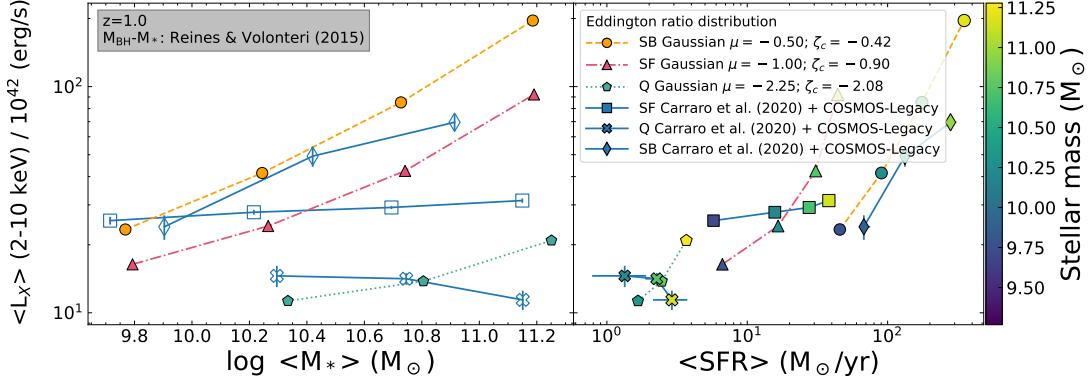


Figure 3.7: L_X as a function of M_* (left) and SFR (right). L_X are obtained at $z = 1.0$ for detected galaxies with a Reines & Volonteri (2015) $M_{\text{BH}} - M_*$ scaling relation and with a Gaussian Eddington ratio distribution as shown in the legend, with a $\sigma = 0.3dex$. In the right panel, SFRs are obtained using the fits from 2.3 for star-forming (SF), quiescent (Q) and starburst (SB) galaxies, and data points are color coded according to M_* . All relations are compared with results from COSMOS data from Chapter 2

redshifts, especially for more massive galaxies. Indeed, continuity equation models clearly show that more massive BHs have formed most of their mass by $z \sim 1$ (e.g., Marconi et al., 2004; Shankar et al., 2020b).

3.2.4 Reproducing the $\langle L_X \rangle - M_*$ relation in starburst, main-sequence and quiescent galaxies

So far we have mostly focused on comparing model predictions with the mean $\langle L_X \rangle - M_*$ relation of star forming main sequence galaxies, which are the vast majority of the detected active galaxies in COSMOS-Legacy. However, AGN activity is routinely detected also in other galaxy life phases. In Figure 2.2 we showed that, at least at $z < 2.25$, starbursts, star forming and quiescent galaxies are characterized by distinct $\langle L_X \rangle - M_*$ relations, which are similar in slope but differ in normalization by a factor of ~ 10 when moving from quiescent galaxies, with the lowest average $\langle L_X \rangle$, to the starbursts, with the highest average $\langle L_X \rangle$ at fixed stellar mass. In the context of our approach, this offset in $\langle L_X \rangle$ at fixed stellar mass could be explained either by a systematic difference in the characteristic Eddington ratio ζ_c and/or by a systematic offset in the normalization of the $M_{\text{BH}} - M_*$ relation, when moving from quiescent to starforming galaxies. In this Section we proceed with a systematic comparison of our

models with the COSMOS-Legacy data focusing on the dependence of the $\langle L_X \rangle - M_*$ relation on galaxy type at fixed redshift, specifically at $z = 1$, though the conclusions we will retrieve below are quite general and can be easily extended to other redshift bins.

In the left panel of Figure 3.7 we explore mocks with a constant input $M_{BH} - M_*$ relation from Reines & Volonteri (2015), but characterized by distinct ζ_c , as labeled (circles, triangles, and pentagons), against the $\langle L_X \rangle - M_*$ relation measured for the three types of galaxies studied in Chapter 2 (blue diamonds, squares and crosses for starbursts, star forming, and quiescent galaxies, respectively). Reproducing the steep increase in mean $\langle L_X \rangle$ at fixed M_* requires, as expected, a proportionally higher value of ζ_c in main sequence and starburst galaxies, assuming the same $M_{BH} - M_*$ relation. We stress that the calculation of the mean $\langle L_X \rangle$ of each galaxy type via Eq. 3.5 would require an additional statistical weight specifying the relative contribution of each galaxy type to the total number of detected active galaxies. As this weight is constant in each stellar mass bin, it would however cancel out when applied to the numerator and denominator of Eq. 3.5. In the right panel of Fig. 3.7 we show the SFRs of the entire sample from Chapter 2 against the luminosity of the X-ray detected sources only. We decided not to use the SFRs from the detected sample since the tracers used for their estimation (IR and UV luminosity) may be polluted by AGN emission and the IR stacking may not achieve a significant signal-to-noise with the low number statistics from this subsample, both leading to non-representative SFRs for these galaxies.

Interestingly, it is apparent from Figure 3.7 that the observed $\langle L_X \rangle - M_*$ relation in starburst galaxies is not a simple power law but tends to show a break that becomes more pronounced in more massive galaxies of mass $\log(M_*/M_\odot) \gtrsim 10.5$ and at lower redshifts. In our modeling, this feature could be naturally reproduced with a further decrease in ζ_c in the most massive galaxies in our sample, which would align with the idea of downsizing, as discussed above. This result supports the view that, already in the early starburst phase, more massive galaxies and their central BHs have accreted their mass earlier and are now in their declining phase, as predicted by some models (e.g., Lapi et al., 2018). We stress that the downsizing in ζ_c would be even more pronounced if steeper $M_{BH} - M_*$ relations were adopted in input. The right panel of Figure 3.7 shows that our chosen values of ζ_c that match the $\langle L_X \rangle - M_*$ relation for each galaxy type also reproduce, at the same time, their respective $L_X - \text{SFR}$ relations, where the SFR is assigned to each galaxy type based on their observed underlying

SFR – M_* relation.

An alternative way to explain the different normalizations of starburst and quiescent galaxies in the $\langle L_X \rangle - M_*$ plane would be to adopt the same ζ_c for all galaxy types and progressively increase the normalization of the $M_{\text{BH}} - M_*$ scaling relation when moving from quiescent to starburst galaxies. We however disfavor such a model. Direct measurements of the $M_{\text{BH}} - M_*$ scaling relation in AGN within a variety of host galaxies (e.g., Reines & Volonteri, 2015; Shankar et al., 2019; Suh et al., 2020, and references therein), have all revealed normalizations that are lower than those typically measured locally in dynamically measured BHs, possibly due to some biases in the latter (e.g., Shankar et al., 2016). In particular, BHs in local elliptical, quiescent galaxies seem to be the most massive ones at fixed host galaxy stellar mass among all samples of local active and normal galaxies (see, for example, Figure 8 in Reines & Volonteri 2015). In addition, also the analysis of the clustering of active, mostly starforming, galaxies at fixed BH mass favors $M_{\text{BH}} - M_*$ scaling relations with a normalization lower than the one measured for local quiescent, early-type galaxies (e.g., Shankar et al., 2020a; Allevato et al., 2021; Viitanen et al., 2021). Direct (or indirect) comprehensive measurements of the $M_{\text{BH}} - M_*$ scaling relation in active starburst galaxies are still unavailable. However, theoretical models suggest that the ratio between BH mass and host galaxy stellar mass in the starburst phase should, if anything, be lower than what observed locally, as the BH grows from a relatively small seed within a host forming stars at exceptionally high rates (see, e.g., Lapi et al., 2014, their Figure 3). More generally, these models suggest that, from an evolutionary point of view, quiescent galaxies should be older galaxies with larger BHs at fixed stellar mass (e.g., Cirasuolo et al., 2005; Granato et al., 2006; Lapi et al., 2006; Shankar et al., 2006; Lapi et al., 2018).

All in all, the evolutionary picture that could be extracted from Figure 3.7 is one in which the central BH and its host galaxy move around a similar $M_{\text{BH}} - M_*$ scaling relation throughout their lifetime. They could start from a main-sequence or even starburst, gas-rich phase, evolving at an almost constant (specific) SFR, as also proposed by theoretical models (e.g. Lapi et al., 2014; Aversa et al., 2015b) and direct observations (Chapter 2, and then gradually switch off their accretion and star formation due to internal gas consumption, thus gradually reducing their SFR and accretion onto the central BH (right panel of Figure 3.7).

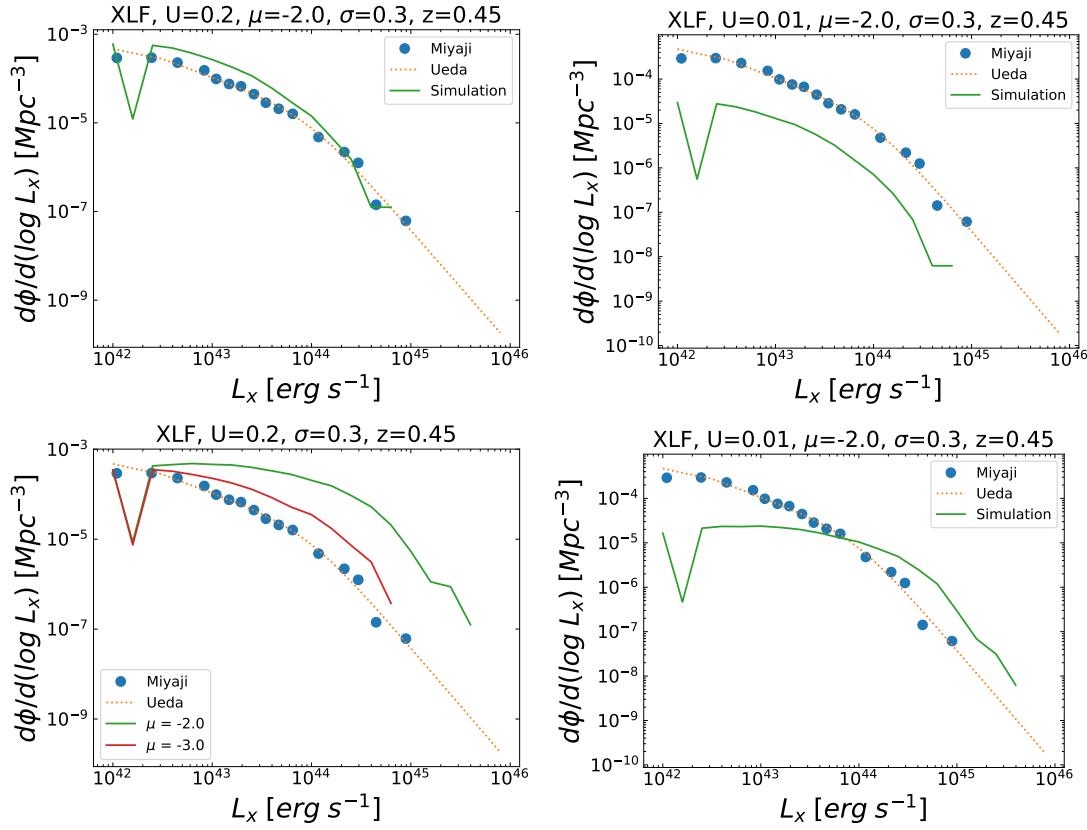


Figure 3.8: XLF from the models using the Eddington ratio better representing the L_X of the data at $z = 0.45$. Top panels: Using Reines & Volonteri (2015) scaling relation, varying the duty cycle $U=0.2$ (left) and $U=0.01$ (right). Bottom panels: Using Sahu et al. (2019) scaling relation, varying the duty cycle $U=0.2$ (left) and $U=0.01$ (right). Models are compared with data from Ueda et al. (2014) and Miyaji et al. (2015) at the same redshift.

3.3 Discussion

We showed in the previous Sections that the mean $\langle L_X \rangle - M_*$ relation of X-ray detected active galaxies is a powerful tool to constrain the mean accretion rate of active BHs ζ_c as a function of time and BH mass, and in ways largely independent of the duty cycle. When coupled to other independent probes, the $\langle L_X \rangle - M_*$ can thus provide an invaluable support in breaking the degeneracies in the accretion parameters of supermassive BHs. For example, as discussed in Section 2, the AGN X-ray luminosity function is a convolution of the underlying BH mass function, which mostly depends on the BH-galaxy scaling relations (e.g., Salucci et al., 1999), the intrinsic fraction of active BHs as a function of BH mass (the duty cycle $U(y, z)$), and the normalized Eddington ratio distribution $P(\log \lambda, z)$ (see, e.g., Shankar et al., 2013, and references therein). Thus, knowledge of the AGN X-ray luminosity function and of the characteristic mean Eddington ratio ζ_c from independent observables, could shed light on the duty cycle, once a robust estimate of the underlying BH-galaxy scaling relation is available from, e.g., AGN clustering measurements (see discussion in Shankar et al., 2020a; Allevato et al., 2021; Viitanen et al., 2021).

Figure 3.8 shows a few examples of the dependencies of the AGN luminosity function on the most relevant model input parameters. We compare the observed X-ray AGN luminosity function³ by Ueda et al. (2014, orange dotted lines) and Miyaji et al. (2015, blue filled circles), with the predictions of our reference model with a constant duty cycle $U = 0.2$, a $M_{\text{BH}} - M_*$ relation from Reines & Volonteri (2015), and a Gaussian $P(\log \lambda, z)$ with $\mu = -2$, a combination able to simultaneously reproduce the observed fraction of X-ray AGN (Figure 3.3) and mean $\langle L_X \rangle - M_*$ relation (Figure 3.4). Despite the crudeness of our model, the top-left panel of Figure 3.8 shows that our reference mock (solid green line) is able to broadly reproduce the data at all luminosities within a factor of $\lesssim 2$, without any extra fine-tuning. On the other hand, when switching to a $M_{\text{BH}} - M_*$ relation with a higher normalization than the one calibrated by Reines & Volonteri (2015), such as the one by Sahu et al. (2019), would tend to significantly overproduce the observed AGN luminosity function, an effect induced by the new $M_{\text{BH}} - M_*$ relation which maps galaxies to more massive BHs and thus more luminous AGN (e.g., Shankar et al., 2020a). To recover the match to the AGN luminosity function with the new $M_{\text{BH}} - M_*$ relation we would require a mean Eddington ratio

³Both luminosity functions do not include Compton-thick sources, thus our duty cycle $U(\log M_{\text{BH}}, z)$ refers to the total fraction of Compton-thin AGN, i.e., those with $\log N_H < 24 \text{ cm}^{-2}$.

ζ_c significantly lower by more than an order of magnitude, as shown in the bottom, left panel (solid, red line), which allows to systematically shift the predicted luminosity function by a factor of $\gtrsim 10$ to fainter X-ray luminosities, in better agreement with the data. Although such a low value of ζ_c could still generate a $\langle L_X \rangle - M_*$ relation in broad agreement with the data, at least at larger stellar masses (by simply proportionally lowering the violet dashed model in the bottom, left panel of Figure 3.4), and also with the observed AGN fraction (pink double dot-dashed line in the bottom right panel of Figure 3.3), it would be inconsistent with independent measurements of the mean Eddington ratios at similar redshifts (e.g., Hickox et al., 2009; Kauffmann & Heckman, 2009; Aird et al., 2019). Alternatively, we could keep the reference value of ζ_c but decrease the duty cycle to $U = 0.01$, as shown in the solid lines reported in the right panels of Figure 3.8. This solution improves the match between the model with higher normalization in the $M_{\text{BH}} - M_*$ relation and the observed AGN luminosity function, at least at the bright end (bottom right panel). However, such a low value of the duty cycle $U = 0.01$ is inconsistent with the much higher fraction of AGN detected in COSMOS-Legacy (Figure 3.3).

Our current work is able to provide additional clues and empirical evidence in support of the (complex) models of supermassive BH evolution in galaxies. According to the standard picture of the early phases of the co-evolution of galaxies and their central BHs (e.g., Granato et al., 2006; Hopkins et al., 2006; Lapi et al., 2018), galaxies undergo a first rapid, gas-rich and strong burst of star formation, during which a (seed) BH can substantially grow at or above the Eddington limit, followed by a more regular and then quiescent phase during which both the star formation and the accretion onto the central BH drop substantially. We already showed in the left panel of Figure 3.7 that, in the context of our modeling, when assuming a constant or slowly varying underlying $M_{\text{BH}} - M_*$ scaling relation, the data tend to favor an evolving characteristic Eddington ratio ζ_c , steadily declining when the galaxy transitions from the starburst to the quiescent phase, and we suggested, based on the comparison with the L_X -SFR relation (right panel of Figure 3.7), that this temporal trend in BH accretion rate should be closely mirrored by the star formation in the host galaxy, in agreement with the expectations from theoretical models. Here we further elaborate on this idea. In the previous Chapter, see, e.g., Figure 2.3, we showed that main-sequence and quiescent galaxies share similar ratios of BHAR and SFR at all probed cosmic epochs, suggesting that the two processes are indeed linked together throughout different galaxy phases. In fact, the mean BHAR/SFR can be written as $\text{BHAR}/\text{SFR} \propto L_{\text{bol}}/\text{SFR} \propto 10^{\zeta_c} M_{\text{BH}}/(kM_*)$,

where $k = \text{SFR}/M_*$ is the specific SFR. Thus, at fixed M_{BH}/M_* , a similar BHAR/SFR ratio as the one observed in star forming and quiescent galaxies, would be induced by a proportional decline in characteristic Eddington ratio ζ_c and specific SFR k within a bin of stellar mass. Analogously, the significantly lower BHAR/SFR in starbursts with respect to quiescent/star forming galaxies, as shown in Fig. 2.3, would be naturally interpreted as a proportionally higher specific SFR k and roughly constant or slightly higher ζ_c in these young gas rich systems, as predicted by some BH evolutionary models (e.g., Lapi et al., 2014; Aversa et al., 2015b).

3.4 Conclusions

In this work we use statistical semi-empirical models to generate accurate mock catalogs of active galaxies, which we analyze in the same manner as in the comparison observational sample from Chapter 2. Our goal is to unveil the input parameters driving the $\langle L_X \rangle - M_*$ relation. We start from a halo mass function at a given redshift, we assign galaxies and BHs to dark matter haloes via the most up-to-date empirical stellar-halo and $M_{\text{BH}} - M_*$ relations, and we assume a SFR depending only on stellar mass and redshift. We explore a range of Eddington ratio distributions $P(\log \lambda, z)$, $M_{\text{BH}} - M_*$ scaling relations and duty cycles $U(\log M_{\text{BH}}, z)$. Our results can be summarized as follows:

- In agreement with previous findings (see, e.g., Aird et al., 2012; Shankar et al., 2013), the apparent increase of AGN detections towards high stellar masses, i.e., the “observed” AGN fraction, is not necessarily caused by AGN being more frequent in more massive galaxies, but we find that it is mostly a consequence of the X-ray survey flux limit, which prevents the detection of the faintest sources with a higher probability of being located in lower mass galaxies.
- The mean $\langle L_X \rangle - M_*$ (or L_X -SFR) relation in detected BHs is largely independent of the AGN duty cycle, but strongly depends on the shape, normalization and scatter of the underlying $M_{\text{BH}} - M_*$ scaling relation and on the characteristic Eddington ratio ζ_c , which play a degenerate role in linking the mean $\langle L_X \rangle$ with the BH mass.
- When assuming a roughly constant $M_{\text{BH}} - M_*$ relation with time, as indicated by many recent observations, current X-ray data on the $\langle L_X \rangle - M_*$ relation favor models with a mean Eddington ratio of a few percent at $z = 0.45$ and rapidly

approaching the Eddington limit at $z \sim 3$, in broad agreement with a variety of independent data sets and theoretical models.

- At fixed redshift $z \gtrsim 1$, the same data sets also show evidence for downsizing, with the most massive BHs having accreted their mass more rapidly than less massive BHs.
- At fixed redshift, the $\langle L_X \rangle - M_*$ relation increases by nearly an order of magnitude in normalization when moving from quiescent to starburst galaxies. Our models suggest that, on the reasonable assumption of a constant $M_{\text{BH}} - M_*$ relation, this increase in mean $\langle L_X \rangle$ is mostly induced by the mean ζ_c being much higher during the starburst, gas-rich phase, and rapidly dropping in the quiescent, gas-poor phase.
- Models consistent with the observed $\langle L_X \rangle - M_*$ relation, independent measurements of the mean Eddington ratios, the observed X-ray AGN fraction, and the X-ray AGN luminosity function, are characterized by input $M_{\text{BH}} - M_*$ relations with normalizations aligned with those of local AGN samples (e.g., Reines & Volonteri, 2015; Shankar et al., 2019), which are often lower than those derived from dynamically measured local BHs.

The main result derived from this work is the evidence that the $\langle L_X \rangle - M_*$ relation can efficiently break degeneracies among input duty cycles, Eddington ratio distributions and also BH-galaxy scaling relations, when the latter are coupled with independent observational probes, such as AGN clustering measurements (Shankar et al., 2020a) and observed AGN fractions, thus representing a powerful test for BH evolutionary models in a cosmological context.

CHAPTER 4

Estimating the gas distribution in AGN nuclei

The work in this Chapter is part of the analysis presented in a paper by Andonie et al., submitted to the Astrophysical Journal.

4.1 Introduction

X-ray emission is a universal characteristic of Active Galactic Nuclei (AGN), thought to arise from inverse Compton upscattering of optical/UV photons of the accretion disk by hot electrons in the corona (e.g., Haardt & Maraschi, 1991). The intrinsic X-ray emission takes a power-law spectral form [$f(E) \propto E^{-\Gamma}$, with typical photon indices of $\langle \Gamma \rangle \sim 1.8\text{--}2.0$; e.g., Nandra & Pounds 1994; Winter et al. 2009; Corral et al. 2011], but can be modified due to interaction with matter in the vicinity of the central Supermassive Black Hole (SMBH). In particular, Compton scattering and photoelectric absorption of the primary X-ray continuum lead to two important features in the X-ray spectrum: the Fe K α emission line and the so-called "Compton-hump". By studying these reprocessed features, together known as the so-called AGN "reflection" component, we can infer the physical properties of the matter from which they originate, and hence probe

the circumnuclear environments of central SMBHs. These physical emissions have to be studied through spectral analysis, as X-ray data do not have enough resolution to allow us to directly observe the very small $\sim pc$ scales where these processes take place, so an indirect approach is needed.

The Fe K α line at 6.4 keV is produced by fluorescence processes related to the absorption of higher energy X-ray photons by neutral Fe atoms. Its spectral profile is generally comprised of broad and narrow components. The narrow component of the Fe K α line (FWHM $\lesssim 10,000 \text{ km s}^{-1}$; e.g., Lubiński & Zdziarski, 2001; Yaqoob & Padmanabhan, 2004; Shu et al., 2010) is an ubiquitous spectral feature of AGN, and in a majority of cases the only component present, while the broad component is harder to pin down since it requires exceptional statistics and broad energy coverage to decouple the line from the underlying continuum and absorption components (e.g., Guainazzi et al., 2006; Marinucci et al., 2014). Nonetheless, when present, reverberation studies suggest that the broad component originates from a compact zone, only a few r_g in extent, around the SMBH (e.g., Cackett et al., 2014), and hence is strongly affected by Doppler and gravitational broadening (e.g., Mushotzky et al., 1995; Tanaka et al., 1995; Yaqoob et al., 1995). On the other hand, the narrow component is thought to be produced somewhere among the outer accretion disk, the broad line region (BLR) and the torus clouds (e.g., Ghisellini et al., 1994; Krolik et al., 1994; Yaqoob et al., 1995), corresponding to light month-to-year distances.

Rapid X-ray continuum variability is commonly observed in unobscured, obscured, and even some heavily obscured AGN and suggests that the primary X-ray emitting source (i.e., the corona) is produced in a compact zone very near to the SMBH (e.g., Mushotzky et al., 1993; De Marco et al., 2013). The X-ray light curve can be analyzed via the power spectral density (PSD) function, which is typically characterized as a power law of the form $P_\nu \propto \nu^\alpha$, where $\nu = 1/T$ is the frequency and α is the power law slope (e.g., Green et al., 1993; Edelson & Nandra, 1999; Vaughan et al., 2003). Typical values for the power law slope in AGN are $\alpha \sim -1$ at lower frequencies, indicative of pink noise, and $\alpha \gtrsim -2$ at higher frequencies, indicative of red noise. The break in the PSD between these two regimes is denoted as $\nu_B = 1/T_B$, and is related to the characteristic X-ray variability timescales of the system.

If we assume that the Fe K α line emission is reprocessed from the same X-ray continuum we observe, we can expect the Fe K α line flux tracks the continuum fluctuations. The light curves of both, primary and reflected components, can still differ by light travel time effects, as the reflected light can travel different and in general longer

paths to the observer. The Fe K α line light curve can therefore be delayed with respect to the continuum and can also be smoothed out, as variations on timescales shorter than the light crossing time of the reflector are damped. The reduction of variability amplitude of the Fe K α line light curve with respect to the continuum, however, can shed light on the size of the reflector, as larger reflectors suppress a larger fraction of the intrinsic variance. The aim of this Chapter is to take advantage of this correlation between primary and reflected component by using temporal considerations and the variability properties to measure the size of the reflector. We did so in a sample of AGN from the work of Andonie et al. (subm.) by simulating X-ray light curves and comparing their reduction in variability amplitude with the observed ones.

4.2 Sample and data

The AGN sample in this Chapter was presented in Carolina Andonie's master thesis. Onto this same sample we are performing an additional work which will be presented in Andonie et al. (in prep.). The sample was chosen to study the spectral and temporal properties of the Fe K α line in AGN, and was selected from the parent input sample the most recent 105-month *Swift*-Burst Alert Telescope (BAT) Survey (Oh et al., 2018), an all-sky survey in the ultra-hard X-ray band (14–195 keV), which provides a relatively unbiased AGN sample at least up to $N_{\text{H}} \gtrsim 10^{24} \text{ cm}^{-2}$ (Ricci et al., 2015). The 105-month Swift-BAT catalog is a uniform hard X-ray all-sky survey with a sensitivity of $8.4 \times 10^{-12} \text{ erg s}^{-1} \text{ cm}^{-2}$ over 90% of the sky in the 14–195 keV band. The survey catalogs 1632 hard X-ray sources, 947 of which are securely classified as AGN but the sample is reduced in order to include only sources with enough multiple observations from either Chandra or XMM. They include one additional target in their sample, the well-known narrow-line Sy1 1H0707–495, which is relatively bright in the 2–8 keV band, yet somehow remains undetected in the BAT 105-month catalog.

Andonie et al. focus on observations in the 2–8 keV band where the Fe K α line is located, and use data from *Chandra* (Weisskopf et al., 2000) and *XMM-Newton* (Jansen et al., 2001) which have a good sensitivity in this band. *XMM-Newton* observations provide better spectral and timing statistics, but suffer from substantial background flaring. On the other hand, *Chandra* observations are more versatile since they offer high spatial resolution to search for extended Fe K α emission on \sim 100-pc to kpc scales, and, when the High Energy Grating is deployed, sufficient spectral resolution to resolve the Fe K α line.

The authors limit their analysis only to those observations for which there is a high likelihood of constraining the Fe K α line, and apply a flux cut of $f_{14-195\text{ keV}} \geq 10^{-11} \text{ erg s}^{-1} \text{ cm}^{-2}$ and/or $f_{2-8\text{ keV}} \geq 4 \times 10^{-12} \text{ erg s}^{-1} \text{ cm}^{-2}$ (these are roughly equivalent for a $\Gamma=1.9$ power law); this resulted in the selection of 252 sources in the local universe ($z < 0.1$), and 28 more distant galaxies with redshifts between 0.1 and 0.56.

We simulate X-ray light curves for those galaxies which have determined PSD function parameters from either Summons (2007), from long term monitoring campaigns performed with the Rossi X-Ray Timing Explorer (RXTE) observatory, or from González-Martín & Vaughan (2012) (GM&V12 hereafter), from *XMM-Newton/pn* observations. For the rest, we estimated the break frequency using the expression

$$\log(T_b) = 1.09 \log(M_{BH}) + -0.24 \log(L_{bol}) - 1.88 \quad (4.1)$$

from GM&V12, where T_b is the PSD bending timescale ($T_b = 1/\nu_b$), M_{BH} is the black hole mass in $10^6 M_\odot$ units, and L_{bol} is the bolometric luminosity in $10^{44} \text{ erg s}^{-1}$ units.

The final sample for which we are able to simulate X-ray light curves is composed of 33 sources which have between 2–68 observations in a time range between $\sim 1600 - 7000$ days. A complete list of the sources, their PSD parameters and observations is reported in Table 4.1.

4.3 Variability definitions

The light curves obtained by Andonie et al. are too sparsely sampled to detect a delay between continuum and Fe line fluctuations, therefore we rely on estimating the reduction in variability amplitude. This can be quantified by calculating the ratio between the excess variance of the continuum (σ_{ct}^2) and Fe K α line (σ_{Fe}^2) light curves, as

$$\xi_r = \frac{\sigma_{Fe}^2}{\sigma_{ct}^2}. \quad (4.2)$$

The normalized excess variance (σ^2) (e.g., Nandra et al., 1997; Paolillo et al., 2004; Papadakis et al., 2008) is a quantitative measurement of the variability amplitude of a light curve, and is defined as

$$\sigma_{rms}^2 = \frac{1}{N_{obs}\bar{x}^2} \sum_{i=1}^{N_{obs}} [(x_i - \bar{x})^2 - \sigma_{err,i}^2] \quad (4.3)$$

while its error due to Poisson noise is

$$err(\sigma_{rms}^2) = \frac{1}{N_{obs}^{3/2}\bar{x}^2} \sum_{i=1}^{N_{obs}} ((x_i - \bar{x})^2 - \sigma_{err,i}^2) - \sigma_{rms}^2 \bar{x}^2. \quad (4.4)$$

The σ_{rms}^2 estimates the intrinsic variance of the light curve, normalized by its mean flux, to produce a dimensionless quantifier that can be easily compared between objects of different brightness or light curves from different energy bands. The last term in Eq. 4.3 estimates the contribution of the observational Poisson noise to the total variance, in order to subtract it, and the error formula in Eq. 4.4 estimates the uncertainty in this subtraction.

The excess variances and their ratios were estimated by running Monte Carlo simulations in order to quantify their variability robustly. These simulations are only aimed at quantifying how the observational Poisson noise affects the estimate of the σ_{rms}^2 . For this purpose we add Gaussian noise to each light curve point, with a σ equal to the error on the flux of each point. In this way, each simulation has the intrinsic variance of the light curve and twice the observational noise and allows to estimate the excess variance and its asymmetric uncertainties for each light curve.

Low intrinsic variances compared to the Poisson noise can sometimes lead to negative values of the σ_{rms}^2 estimate, since the uncertainty in the Poisson noise can be larger than the difference in Eq. 4.3. We will consider light curves as significantly variable if the lower 16% bound of the excess variance distribution is positive. Table 4.1 shows the median and the 16% and 84% bounds of the ratio distributions of the excess variance of each light curve. These data are also plotted in Fig. 4.1.

Significant variability (i.e. positive lower bound on the σ_{rms}^2) is detected in the continuum of all objects except for 4C+29.30. Significant Fe K α line variability is detected in 2MASXJ23444387, 4C+74.26, Cen A, Circinus galaxy, IC4329A, MR2251-178, MRK 1040, MRK 1210, MRK 3, NGC 1068, NGC 1275, NGC 1365, NGC 253, NGC 2992, NGC 3783, NGC 4151, NGC 4388, NGC 6300 and NGC 7582. In many objects, however, the variability of the Fe K α line is consistent with the observational noise, within its uncertainties. This is the case for all objects for which the lower bound on the Fe K α line excess variance is negative. The upper bounds on the Fe K α line variability in those objects is still of interest, depending on how this bound compares to the variance of the continuum.

If the Fe K α line flux tracks the fluctuations of the continuum flux, then we expect their variances to be related. The ratio of the variances ξ should be similar to 1 if the reflector is small compared to the timescale of the fluctuations, and smaller than 1 if the reflector is large. Therefore, upper bounds smaller than 1 in the ratio column of Table 4.1 can allow us to place a lower limit on the size of the reflector. This is true for example in the first two objects in the table. Even though their Fe K α line variance

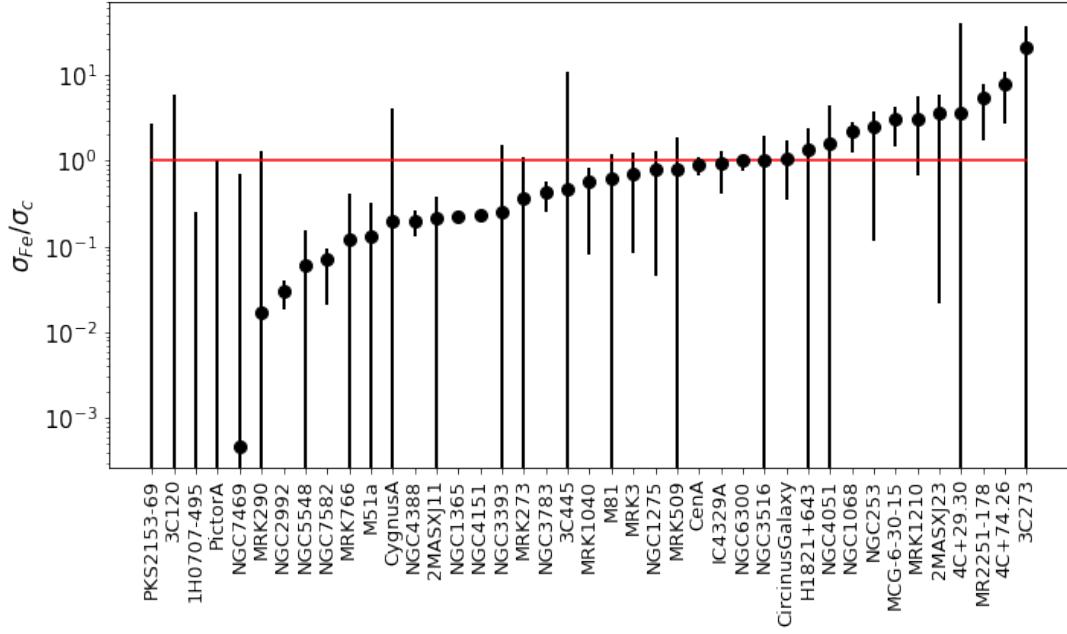


Figure 4.1: The ratio between the normalized excess variance of the Fe K α line and continuum light curves $\xi = \sigma_{Fe}/\sigma_{ct}$. The error bars correspond to the 16% and 84% bounds of the normalized excess variance distributions.

is not significant, its upper limit is so far below the variance of the continuum that a lower limit can be placed on the size of the reflector, because if the reflector was any smaller, the Fe K α line variability would be detectable. Conversely, lower bounds of the variance ratio $\xi > 0$ can put upper limits on the reflector size. One or both of these limits are therefore measurable in many objects in the table. We describe this analysis below in Sec. 4.4.

Source	$\xi(50\%)^{84\%}_{16\%}$	$\log M_{BH}$ (M_\odot)	$\log L_{bol}$ (erg/s)	ν_b (Hz)	ν_b (days $^{-1}$)	α_h	Ref.	N Obs	Duration (days)
1H0707-495	-0.18 $^{0.28}_{-0.82}$	-	-	$3.98 \cdot 10^{-4}$	34.	2.4	GM&V12	14	6929
2MASXJ11315154 -1231587	0.20 $^{0.39}_{-0.026}$	8.81	46.6	$6.22 \cdot 10^{-7}$	0.054	-	Eq. 4.1	41	4990
3C120	-1.3 $^{6.6}_{-19.}$	7.74	45.2	$7.83 \cdot 10^{-6}$	0.68	-	Eq. 4.1	11	4879
3C273	18 $^{36.}_{-17.}$	8.84	47.0	$7.30 \cdot 10^{-7}$	0.063	-	Eq. 4.1	47	6595
4C+29.30	3.8 $^{40.}_{-27.}$	8.28	44.9	$1.28 \cdot 10^{-6}$	0.11	-	Eq. 4.1	6	3245
4C+74.26	7.7 $^{11.}_{-2.1}$	9.83	46.0	$2.00 \cdot 10^{-8}$	0.0017	-	Eq. 4.1	6	1571
CenA	0.92 $^{1.1}_{-0.66}$	7.77	43.1	$2.26 \cdot 10^{-6}$	0.20	-	Eq. 4.1	39	6581
CircinusGalaxy	1.1 $^{1.8}_{-0.33}$	6.23	43.5	$1.07 \cdot 10^{-4}$	9.3	2.1	GM&V12	18	5006

Continued on next page

4.3. VARIABILITY DEFINITIONS

Source	$\xi(50\%)^{84\%}_{16\%}$	$\log M_{\text{BH}}$ (M_{\odot})	$\log L_{bol}$ (erg/s)	ν_b (Hz)	ν_b (days $^{-1}$)	α_h	Ref.	N Obs	Duration (days)
CygnusA	$-0.51^{3.7}_{-6.0}$	9.43	45.6	$5.30 \cdot 10^{-8}$	0.0046	-	Eq. 4.1	68	6209
IC4329A	$0.99^{1.3}_{0.47}$	7.81	45.0	$5.85 \cdot 10^{-6}$	0.51	-	Eq. 4.1	13	6222
M81	$0.75^{1.2}_{-0.24}$	7.90	39.5	$2.13 \cdot 10^{-7}$	0.018	-	Eq. 4.1	38	6157
MCG-6-30-15	$3.0^{4.2}_{1.5}$	6.14	43.9	$3.80 \cdot 10^{-5}$	3.3	1.9	Summons (2007)	12	4589
MR2251-178	$5.3^{7.9}_{2.0}$	8.20	45.8	$2.50 \cdot 10^{-7}$	0.022	2.5	Summons (2007)	12	5489
MRK1040	$0.57^{0.82}_{0.18}$	7.41	44.6	$1.55 \cdot 10^{-5}$	1.3	-	Eq. 4.1	8	4770
MRK1210	$3.4^{6.1}_{0.90}$	6.76	44.3	$9.94 \cdot 10^{-5}$	8.6	-	Eq. 4.1	7	2496
MRK273	$0.43^{1.0}_{-0.90}$	8.78	44.1	$1.78 \cdot 10^{-7}$	0.015	-	Eq. 4.1	8	6149
MRK290	$0.13^{1.4}_{-1.7}$	7.28	44.4	$2.06 \cdot 10^{-5}$	1.8	-	Eq. 4.1	8	1042
MRK3	$0.73^{1.3}_{0.12}$	6.72	44.8	$1.51 \cdot 10^{-4}$	13.	-	Eq. 4.1	19	5511
MRK509	$0.76^{1.8}_{-0.81}$	8.05	45.3	$7.60 \cdot 10^{-8}$	0.0066	1.5	Summons (2007)	18	4335
MRK766	$0.11^{0.43}_{-0.35}$	6.60	43.9	$2.90 \cdot 10^{-4}$	25.	2.9	Summons (2007)	11	5172
NGC1068	$2.3^{2.8}_{1.4}$	6.93	43.9	$4.81 \cdot 10^{-5}$	4.2	-	Eq. 4.1	12	5301
NGC1275	$0.82^{1.3}_{0.025}$	7.55	45.1	$1.38 \cdot 10^{-5}$	1.2	-	Eq. 4.1	44	6650
NGC1365	$0.22^{0.25}_{0.19}$	7.60	43.4	$2.21 \cdot 10^{-6}$	0.19	-	Eq. 4.1	9	3314
NGC2992	$0.031^{0.040}_{0.019}$	8.33	43.1	$4.09 \cdot 10^{-7}$	0.035	-	Eq. 4.1	13	3644
NGC3393	$0.29^{1.7}_{-1.7}$	7.52	43.8	$7.24 \cdot 10^{-6}$	0.63	-	Eq. 4.1	5	3199
NGC3516	$1.0^{2.0}_{-0.68}$	7.39	43.9	$6.60 \cdot 10^{-6}$	0.57	2.9	Summons (2007)	9	2204
NGC3783	$0.45^{0.58}_{0.29}$	7.37	44.6	$1.30 \cdot 10^{-5}$	1.1	2.6	Summons (2007)	13	6179
NGC4051	$1.8^{4.4}_{-3.3}$	6.13	42.4	$5.10 \cdot 10^{-4}$	44.	2.5	Summons (2007)	42	5866
NGC4151	$0.24^{0.25}_{0.22}$	7.56	43.4	$2.60 \cdot 10^{-7}$	0.022	2.2	Summons (2007)	35	5769
NGC4388	$0.20^{0.26}_{0.13}$	6.94	44.2	$5.45 \cdot 10^{-5}$	4.7	-	Eq. 4.1	5	3267
NGC5548	$0.042^{0.15}_{-0.071}$	7.72	44.3	$1.30 \cdot 10^{-6}$	0.11	3.5	Summons (2007)	21	5823
NGC6300	$1.0^{1.1}_{1.0}$	6.57	43.0	$8.84 \cdot 10^{-5}$	7.6	-	Eq. 4.1	6	3026
NGC7469	$-0.036^{0.68}_{-1.2}$	6.96	44.4	$5.60 \cdot 10^{-5}$	4.8	-	Eq. 4.1	12	5480
NGC7582	$0.070^{0.094}_{0.018}$	7.74	44.7	$5.89 \cdot 10^{-6}$	0.51	-	Eq. 4.1	7	6371
Pictor A	$-0.34^{0.88}_{-3.1}$	6.80	44.6	$1.06 \cdot 10^{-4}$	9.1	-	Eq. 4.1	17	5471

Table 4.1: The galaxy sample from Andonie et al. (in prep.), its observations and its PSD parameters. Median ξ parameters are reported with the 16% to 84% bounds of the distributions of normalized excess variance. Black hole masses and L_{bol} are from BASS DR2 (in prep.). ν_b and α_h (if available) are shown with the work they was extracted from and, whenever not available, the ν_b estimation with Eq. 4.1 is shown.

The longer the light crossing time of the reflector, the smaller the expected value of ξ_r is. Although it is straightforward to estimate ξ_r as a function of the break frequency of the continuum light curve power spectrum and the light crossing time of the reflector, we expect a large scatter in measured values of ξ_r due to the small number of data points and the stochastic nature of the light curves. We therefore took the Monte Carlo

approach described below to place meaningful constraints on the size of the reflector.

4.4 Light curves simulation

We simulate light curves following the method of Timmer & Koenig (1995), which consists in simulating white noise in Fourier space, defined as a normally distributed process with a real and imaginary part with a standard deviation that depends on the filter function, our power spectrum. We assume a power spectrum with the shape of a bending power law in Fourier space:

$$PS = \frac{A\nu^{-\alpha_L}}{1 + (\nu/\nu_b)^{\alpha_H - \alpha_L}} \quad (4.5)$$

with a low-frequency slope $\alpha_L = 1$ bending at higher frequencies to a steeper slope α_h at a characteristic timescale or break frequency ν_b . The normalization was given a value of $A = 0.001$ which conforms to the majority of sources measured in Summons (2007). The variance scales linearly with the normalization A , so varying A will shift the mean expected variance and its scatter by the same amount. Differences in A by a factor of 2 (higher or lower) are consistent with the monitored sample. When no estimate of a high frequency slope is available from the literature, we assumed $\alpha_H = 2$.

To simulate the corresponding reflected light curves we multiplied the simulated continuum light curve in Fourier space for a $\text{sinc}(\nu\tau)$ function, which corresponds to a top hat filter, and shifted the inverse Fourier transformed curve forward in time by $\tau/2$ days. This simple setup corresponds to the reflection by a spherical thin shell of radius $R = c\tau/2$ and can be viewed as an immediate response of the front end of the reflector, followed by the reflection of the rest of the shell until the light from the back end finally reaches the observer $2R/c = \tau$ days after the start of the response. This particular response function was chosen for simplicity and reproduces the main characteristics of a reflected light curve, which is sufficient to estimate the size, though not the geometry, of the reflector.

For each source we consider its spectral parameters and in particular we choose a sampling time-scale dt based on its bending time-scale as $dt = 10^{\text{floor}(\log_{10}(T_b/100))}$, i.e. the sampling resolution is at least a 100th of the T_b scale and is an entire number in the representation $\log(dt)$. This choice results in values $-4 \leq \log(dt) \leq -1$. We resampled the simulated light curves according to the observed epochs and added Gaussian noise to each simulated data-point with a standard deviation scaled based on the error of the observation at that epoch, in order to reproduce the additional variance of the

4.4. LIGHT CURVES SIMULATION

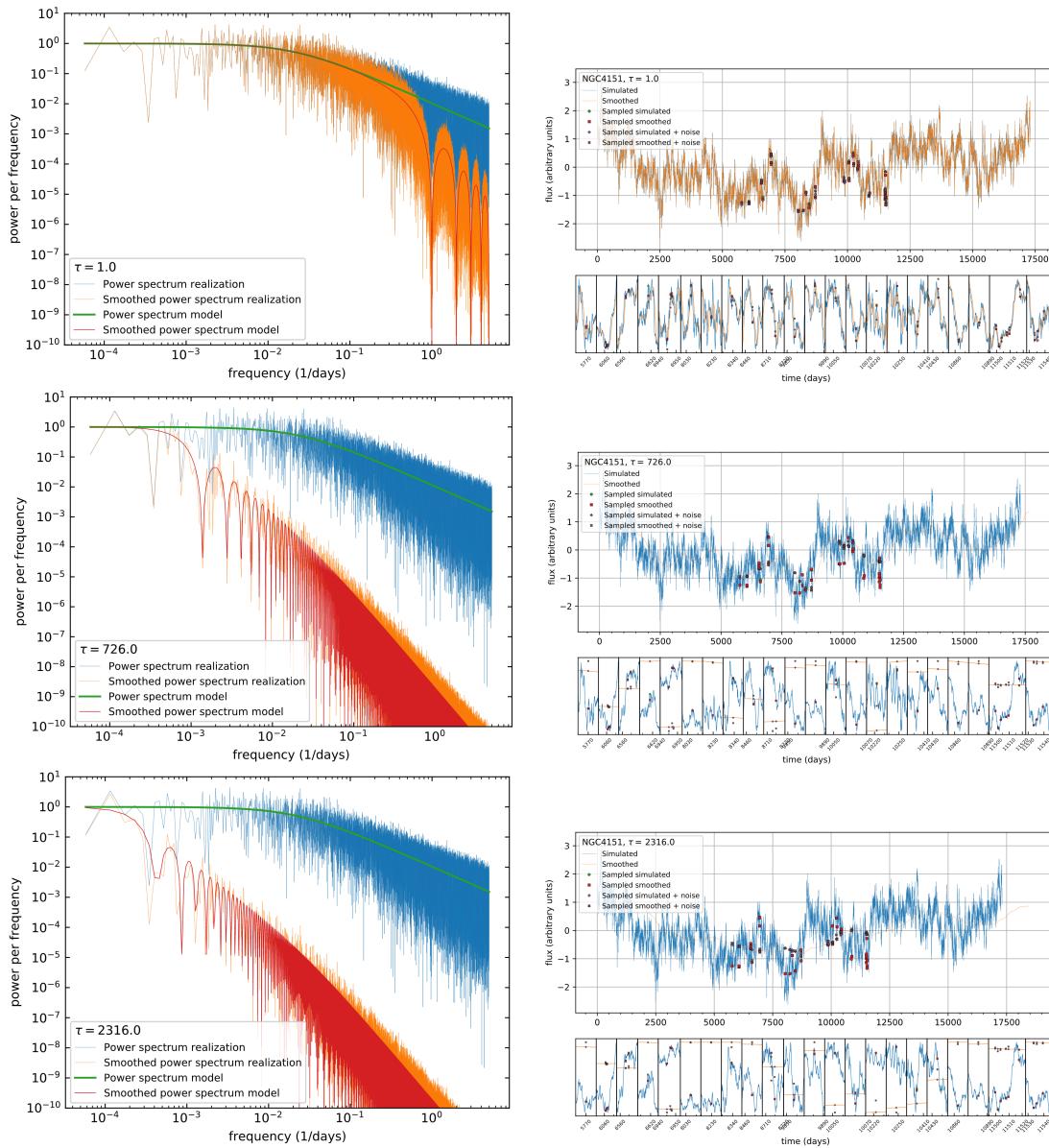


Figure 4.2: Power spectrum NGC4151

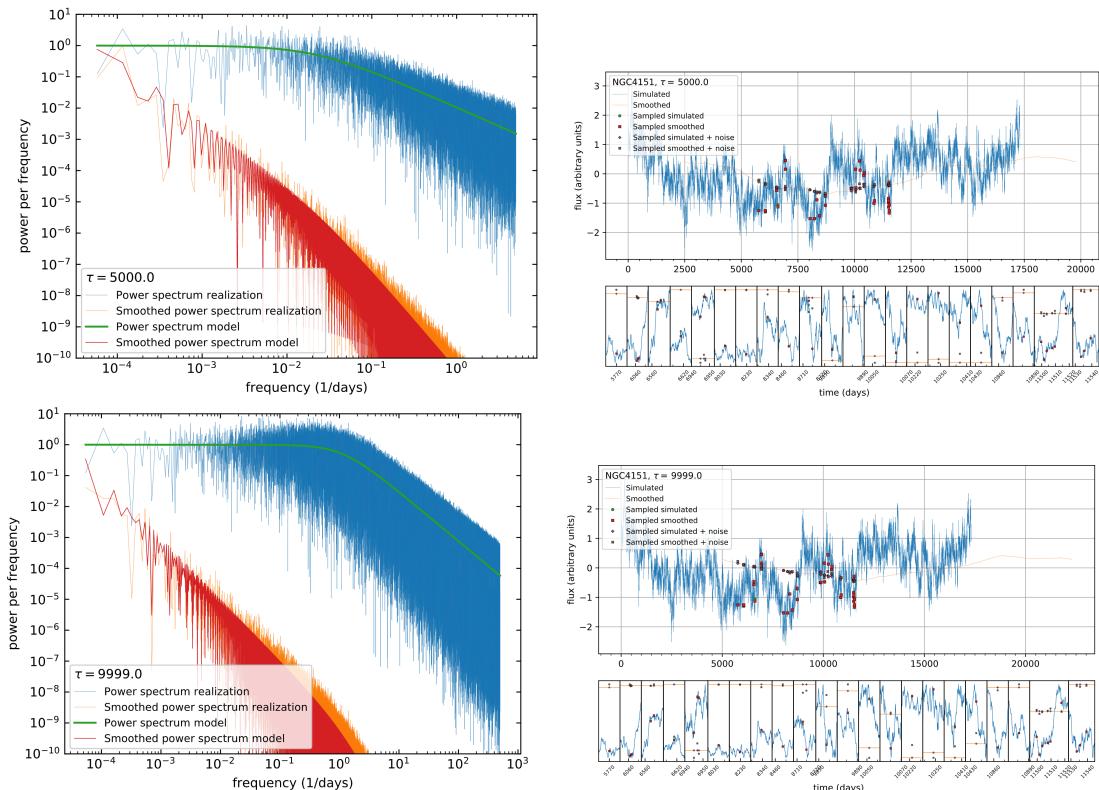


Figure 4.3: Power spectrum NGC4151

data. Finally, we computed the ratio of the variances ($\xi_{r,\text{sim}} = \sigma_{\text{Fe},\text{sim}}^2 / \sigma_{\text{c},\text{sim}}^2$). As expected, this ratio decreases as the light crossing time of the reflector increases. We ran sets of 50 simulations and recorded the mean and root-mean-square deviations (rms) of $\xi_{r,\text{sim}}$ for a range of values of the light crossing time. We implemented the following statistic to find the τ values which correspond to our estimated reprocessor size and its upper and lower limits:

$$X(\tau) = \frac{\xi_{\text{data}} - \langle \xi_{\text{sim}} \rangle}{(\xi_{\text{sim}})^{\text{rms}}} \quad (4.6)$$

where $\xi_{r,\text{real}}$ is the median value of the ξ_r distribution.

Examples of the power spectra used to simulate light curves are reported in the left panels of Figs. 4.5 and 4.5 for NGC 3783 and Figs. 4.2 and 4.3 for NGC 4151, for both the continuum and Fe K α ("smoothed"). These two galaxies were chosen as an example of very different ν_b frequencies and because we were able to find all three τ values corresponding to $X = 0, 1, -1$. In the model power spectrum (green line) we can see a decrease in power after the break frequency, which is $\nu_b = 1.1 \text{ days}^{-1}$ for NGC 3783 and $\nu_b = 0.022 \text{ days}^{-1}$ for NGC 4151. We also see a decrease in the smoothed power spectrum (red line) at frequencies higher than $1/\tau$, where τ is shown in the legend title of each figure and is shown in increasing order.

In the right panels of the aforementioned Figures we show the light curves associated with the power spectra on the left, and we also show the resampled data points from the light curves. The blue light curve corresponds to the realization of the power spectrum in blue as well in the left panel, and the orange light curve is the smoothed curve which corresponds to the smoothed power spectrum realization, in orange as well. It can be seen that a higher τ corresponds to a higher smoothing in the light curve, as expected, and a non obvious correlation between the two light curves, especially with such poor sampling. This can be better seen in the lower panels on the right, which are a zoom in correspondence to the observed data points, but especially in Fig. 4.6 and 4.4, which show flux-flux plots for some τ values.

The τ^* values that return $X = 0, 1, -1$ correspond to the reprocessor size in light days and its upper and lower limits at $\pm 1\sigma$, respectively. In order to constrain the size of the reprocessor, we first explore a range of τ values between 1-10,000 days, and then numerically look for the roots $X = 0, 1, -1$ using Newton's method. The resulting $X(\tau)$ relations explored for each source are shown in Appendix B.1 while the reprocessor sizes thus determined are discussed in Section 4.5.

4.5 Correlation between Fe line and continuum flux

A strong, positive correlation between these two fluxes, which is much simpler to assess compared to more complicated lag analyses, should indicate that the reflector lies in close proximity to the source of the X-ray continuum emission. Thus we explore potential correlations between the observed 2–10 keV continuum and Fe K α line light curves for sources in our sample. As mentioned above, any lag in the Fe K α line light curves, which is expected from reprocessing travel time delays, can reduce the apparent correlation between the observed light curves. The degree of loss of correlation is a function of the ratio between the light crossing time of the reprocessor and the characteristic timescales of fluctuations, as well as the geometry of the reprocessor.

The green data-points and the relative green best fit in Figs. 4.6, 4.4 show one realization of the resampled continuum and reprocessed (Fe K α) light curves corresponding to our best estimate of the reprocessor size τ^* while in orange and red we show the resampled data points and best fit our our upper and lower limit (one realization). These relations don't necessarily show a correlation between the observed data (shown with brown crosses), and simulated light curves, but this is due to the delay and the sampling of the data, as we've shown in the previous paragraph, and it doesn't imply that they aren't correlated.

To better show this we refer the reader to Figs. 4.7, 4.8 4.9 4.10, 4.11, where for each source in our sample for which we were able to determine τ^* such that $X(\tau^*) = 0$, we show all the best fits of the flux-flux relations of the 50 simulated light curves, versus the data and their best fit, performed with weighted least squares assuming errors on the Fe K α line only, as the uncertainty on the Fe K α line light curve is generally much higher than in the continuum. For each source we show in the bottom left box the Pearson's correlation coefficient and the Spearman's rank of the observed data, and the reduced χ^2 (or $\tilde{\chi}^2$) of the fit. The Pearson's correlation coefficient is a measure of linear correlation between the two variables, while the Spearman's rank assesses how well the relationship between the two variables can be described using a monotonic function, and both assume a value of 1 for complete correlation/increasing monotonicity, 0 for no correlation and -1 for complete anti-correlation/decreasing monotonicity. The $\tilde{\chi}^2$ tells us the quality of the description of the linear fit, so a $\tilde{\chi}^2 \approx 1$ assesses a good description of the data by the fit, $\tilde{\chi}^2 > 1$ indicates that the fit has not fully captured the data, $\tilde{\chi}^2 \gg 1$ indicates a poor model fit, and $\tilde{\chi}^2 < 1$ indicates that the model is “over-fitting” the data: either the model is improperly fitting noise, or

the error variance has been overestimated. In most cases we see values ranging from a decent correlation ($\rho > 0.5$) and $\tilde{\chi}^2 > 1$ to a poor correlation with $0 < \rho < 0.5$ and $\tilde{\chi}^2 \gg 1$. We also show in the right panels a histogram of the slopes of the linear fits of the simulations versus the slope of the observed data. We can see that in some cases the slope of the data is compatible with the range of the slopes from the simulated light curves.

We show in Table 4.2 the τ^* values of each source sorted by the median ξ . As already mentioned in Section 4.3, per definition in Eq. 4.2, the ξ will be ≈ 1 if the reprocessor is nuclear, i.e. very small, and $\xi \gtrsim 0$ if the reprocessor is extended. There are also two cases in which the size of the reprocessor cannot be constrained, these are $\xi < 0$ when the uncertainty on the data is greater than their variation, and $\xi > 1$ if the variance of the Fe K α line is greater than that of the continuum. The first case, $\xi < 0$, occurs when the uncertainty on the data is greater than their variation and this happens to be the case for the Fe K α light curve of the affected sources. However, the upper limit on the ratio of excess variance of these sources is still > 0 , therefore it could also be a matter of overestimation of the noise because of the error on the noise (Eq. 4.4). In the latter case, $\xi > 1$, either the poor sampling of the light curves did not seize the wider variations of the continuum, or the Fe K α line may not be reflecting the variations of the continuum and therefore the reprocessor cannot be modeled in our physical description of the AGN as introduced in Sec. 4.4. A physical example of this situation could be a radio loud source where a cloud reflects towards us the beamed X-ray emission associated with the powerful jets, leading to stronger and/or more rapid variations in the Fe K α line than in the continuum. We see in Table 4.2 that we were not able to constrain the τ^* values when $\xi < 0$ and $\xi > 1$, mainly because data quality was poor. Instead, in the intermediate cases we were able to place some constraints, especially when both the upper and lower limits of ξ are well constrained in the range (0,1), like in the cases of NGC 2992, NGC 7582, NGC 4388, NGC 1365, NGC 4151, NGC 3783, MRK 1040. When ξ and its lower limit were very low and/or compatible with negative, meaning that the reprocessor is very extended, we were able to determine a lower limit on the size of the reprocessor. On the contrary, when ξ and its upper limit were close to one, meaning that the reprocessor is small, we were able to place an upper limit on the size of the reprocessor. Roughly $\approx 22\%$ (9/41) lie above 1, although all are consistent with unity (red line in Fig. 4.1) to within 2σ uncertainties. In total, more than 50% have values consistent with unity, implying little damping of the continuum by the reflector (i.e., angle-averaged light-crossing timescales to the reflector

comparable to the continuum variability timescales), or alternatively that we are not observing the true continuum fluctuations. Sometimes, when the ξ value was not well constrained and outside of the (0,1) range we were able to determine the size of the reprocessor, but not its 1σ limits, or maybe we weren't able to determine any constraint. The complete range of $X(\tau)$ determined for each source is shown in Appendix B.1.

We were able to completely constrain the size of the reprocessor, i.e. best value size and 1σ uncertainties, for a total of 4 sources and at least one of the limits for 13 sources. For the remaining 16 sources we weren't able to constrain the reprocessor size. This is still a remarkable result, given the sparsity of our light curves and the inconsistent quality of the archival data. Furthermore, the physical model we adopt is extremely simple and may not properly describe all of the source. In fact, more realistic distributions could be thicker, clumpier, and have a toroidal or ionization cone-like structure seen at a specific orientation with respect to the line-of-sight, all of which can impact delay times and smoothing.

	$\xi(50\%)^{84\%}_{16\%}$	τ_{inf} (light days)	τ_{best} (light days)	τ_{sup} (light days)
3C120	$-1.3^{6.6}_{-19.}$	-	-	-
CygnusA	$-0.51^{3.7}_{-6.0}$	-	-	-
PictorA	$-0.34^{0.88}_{-3.1}$	-	-	-
1H0707-495	$-0.18^{0.28}_{-0.82}$	1	-	-
NGC7469	$-0.036^{0.68}_{-1.2}$	-	-	-
NGC2992	$0.031^{0.040}_{0.019}$	377	9713	-
NGC5548	$0.042^{0.15}_{-0.071}$	99	-	-
NGC7582	$0.070^{0.094}_{0.018}$	200	8548	-
MRK766	$0.11^{0.43}_{-0.35}$	-	18	-
MRK290	$0.13^{1.4}_{-1.7}$	-	231	-
2MASXJ11315154-1231587	$0.20^{0.39}_{-0.026}$	112	3168	-
NGC4388	$0.20^{0.26}_{0.13}$	2	3013	6602
NGC1365	$0.22^{0.25}_{0.19}$	45	1999	4652
NGC4151	$0.24^{0.25}_{0.22}$	726	2316	5000
NGC3393	$0.29^{1.7}_{-1.7}$	-	-	-

Continued on next page

	$\xi(50\%)_{16\%}^{84\%}$	τ_{inf} (light days)	τ_{best} (light days)	τ_{sup} (light days)
MRK273	$0.43_{-0.90}^{1.0}$	-	1364	5703
NGC3783	$0.45_{-0.29}^{0.58}$	6	271	1675
MRK1040	$0.57_{-0.18}^{0.82}$	-	14	903
MRK3	$0.73_{-0.12}^{1.3}$	-	-	3
MRK509	$0.76_{-0.81}^{1.8}$	-	84	2418
NGC1275	$0.82_{-0.025}^{1.3}$	-	15	-
CenA	$0.92_{-0.66}^{1.1}$	-	-	51
IC4329A	$0.99_{-0.47}^{1.3}$	-	1	385
NGC3516	$1.0_{-0.68}^{2.0}$	-	-	1988
CircinusGalaxy	$1.1_{-0.33}^{1.8}$	-	-	5
NGC4051	$1.8_{-3.3}^{4.4}$	-	-	-
NGC1068	$2.3_{-1.4}^{2.8}$	-	-	-
MCG-6-30-15	$3.0_{-1.5}^{4.2}$	-	-	-
MRK1210	$3.4_{-0.90}^{6.1}$	-	-	-
4C+29.30	$3.8_{-27.}^{40.}$	-	-	-
MR2251-178	$5.3_{-2.0}^{7.9}$	-	-	-
4C+74.26	$7.7_{-2.1}^{11.}$	-	-	-
3C273	$18_{-17.}^{36.}$	-	-	-

Table 4.2: For each galaxy we show the τ^* such that $X(\tau_{inf}) = -1$, $X(\tau_{best}) = -1$, and $X(\tau_{sup}) = -1$. We also report the ξ values previously shown in Table 4.1, and we order the Table based on this column for clarity reasons.

4.6 Discussion

After constraining the size of the reprocessor for a portion of the galaxies in our sample, we want to find a relation with AGN properties. A possibility is for the reprocessor size to scale with the mass of the black hole M_{BH} . In fact, the size of the circum-nuclear environment around the black hole scales, at least partially, with the M_{BH} , e.g. the Schwarzschild radius, the sphere of influence, internal dust radius, the broad line

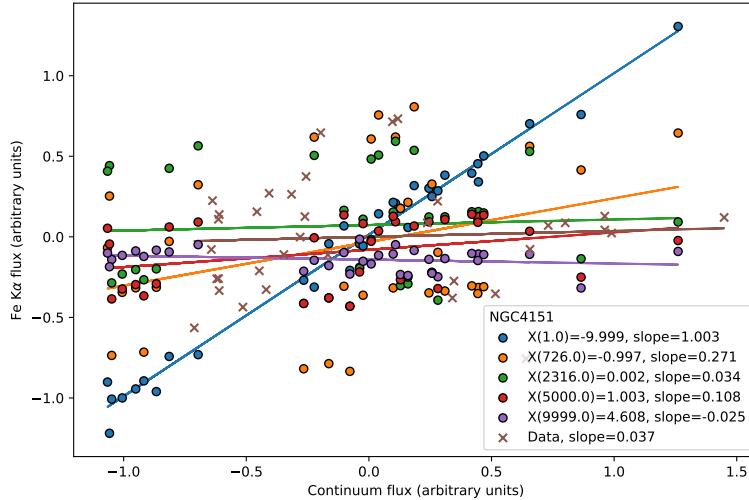


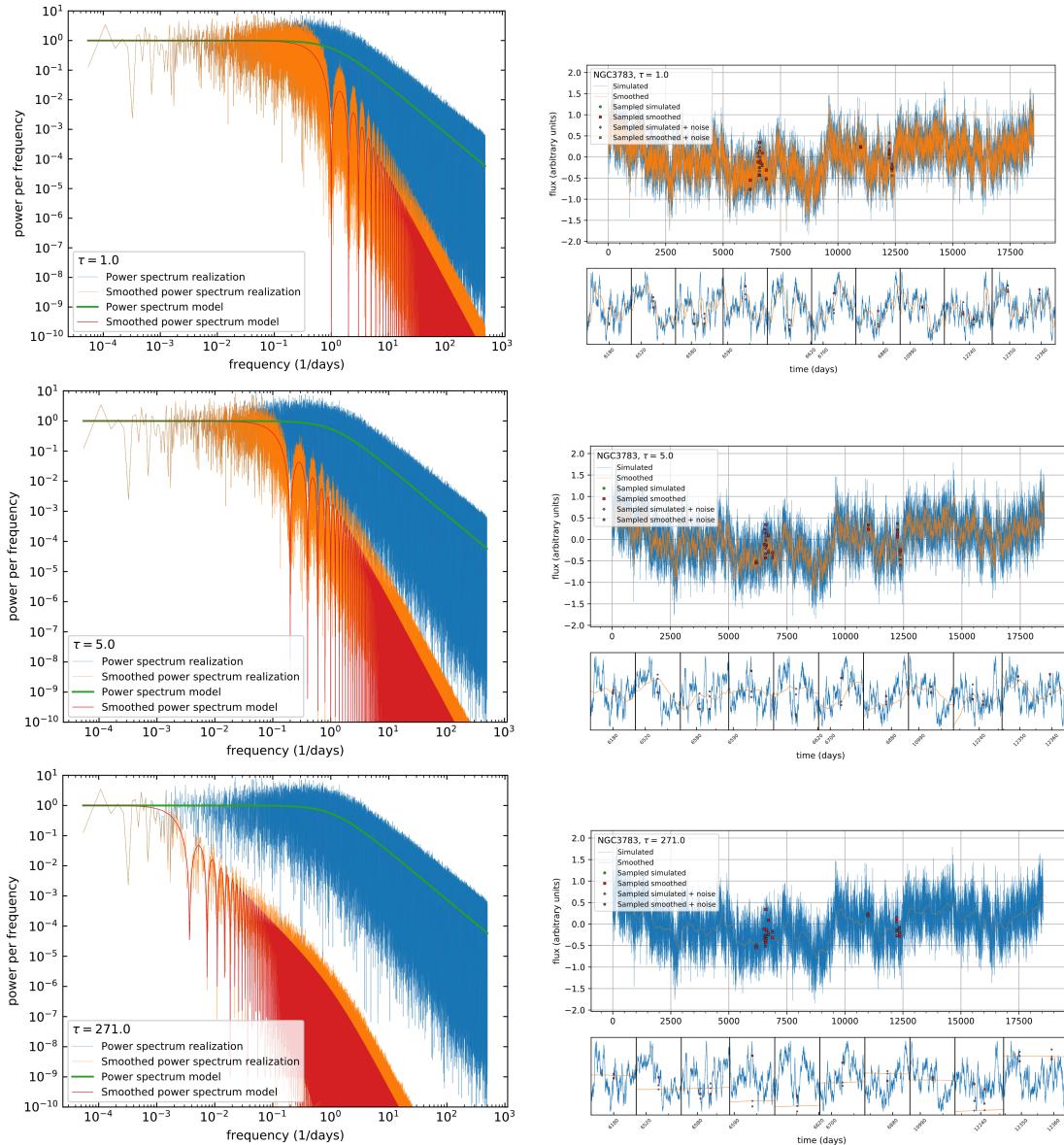
Figure 4.4: Flux - flux plot NGC4151

region, so it seems viable for the material reprocessing the X-ray emission from the corona to increase its size with M_{BH} .

In Figure 4.12 we show all the reprocessor sizes in light day units, with their confidence ranges, as a function of M_{BH} (we remind the reader that the diameter of the reprocessor corresponds to $d = c\tau$). Since we are considering only a range of 1-10,000 light days in size, for the cases where we couldn't constrain the 1σ ranges (see sources with missing τ_{inf} and/or τ_{sup} on Table 4.2) we are showing error-bars covering from the best size τ^* to the lower/upper extreme of the τ range. Along with these we also show a log-log best fit to the data which shows a slope 0.66 ± 0.53 . With the help of the Pearson's correlation coefficient, the Spearman's rank and the χ^2 we see that there is a mild positive correlation in this relation with a large scatter of $\sim 1dex$.

This seemingly weak correlation though, is an interesting starting point that poses this analysis method as a promising one. In fact it allows to place constraints on the size of the reprocessor of quite a few sources and sets a good base for future monitoring campaigns. In fact, by placing lower limits on the size of the reprocessor we can define good sampling frequencies and by placing upper limits we can define a convenient duration of the monitoring campaign in order to measure the lag between the two light curves. Whenever an upper limit is not present, or when this is $\gtrapprox 10^3$, it suggests that our kind of analysis based on the measure of the smoothing in the Fe K α light curve, is still the best way to go, as very long monitoring campaigns are time expensive and would require years or even decades to lead to a lag measurement. Re-

4.6. DISCUSSION



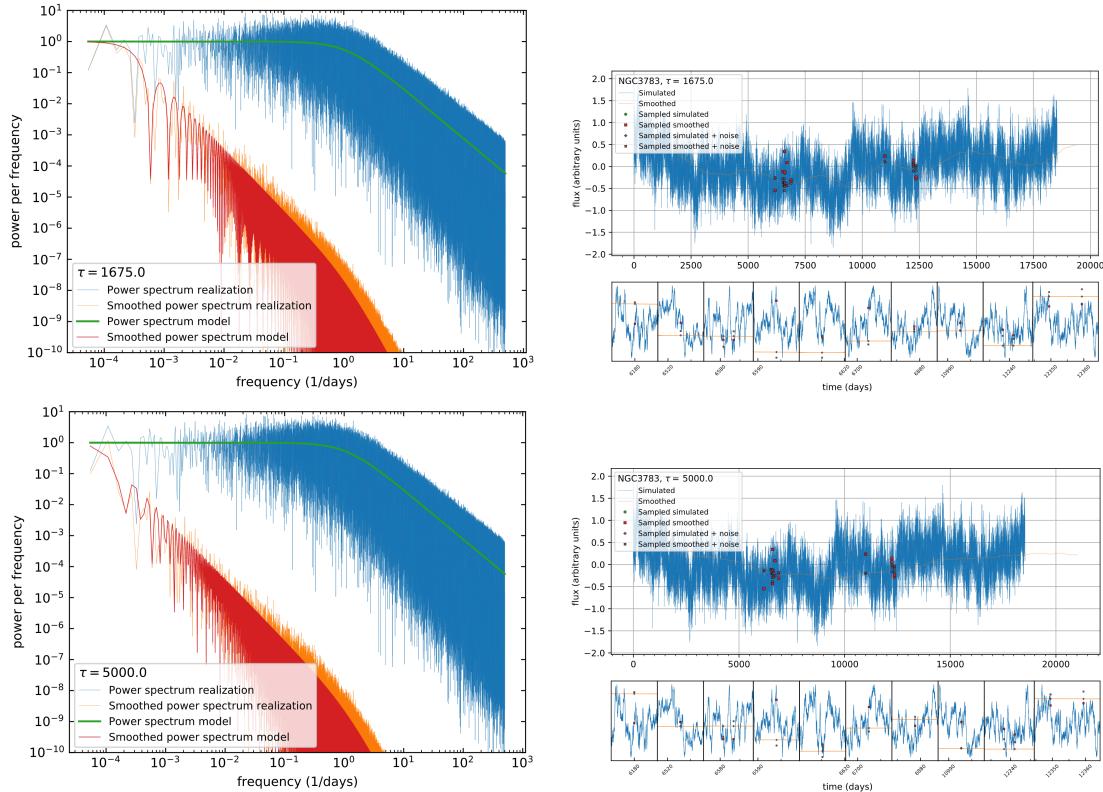


Figure 4.5: Power spectrum NGC 3783

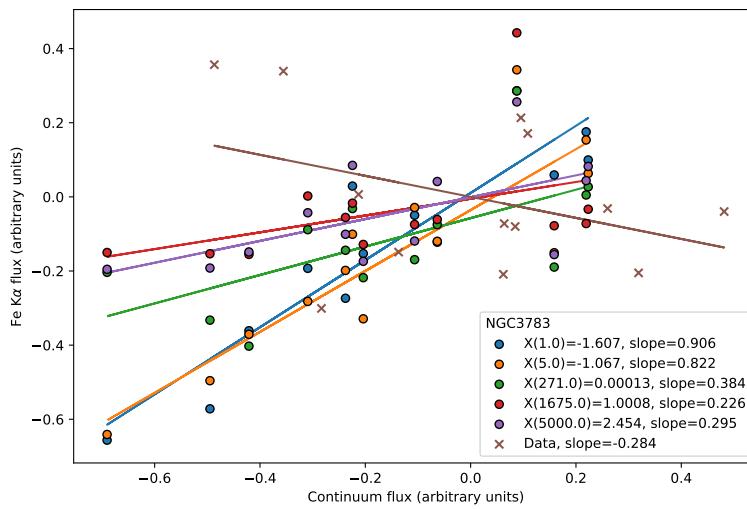


Figure 4.6: Flux - flux plot NGC 3783

4.6. DISCUSSION

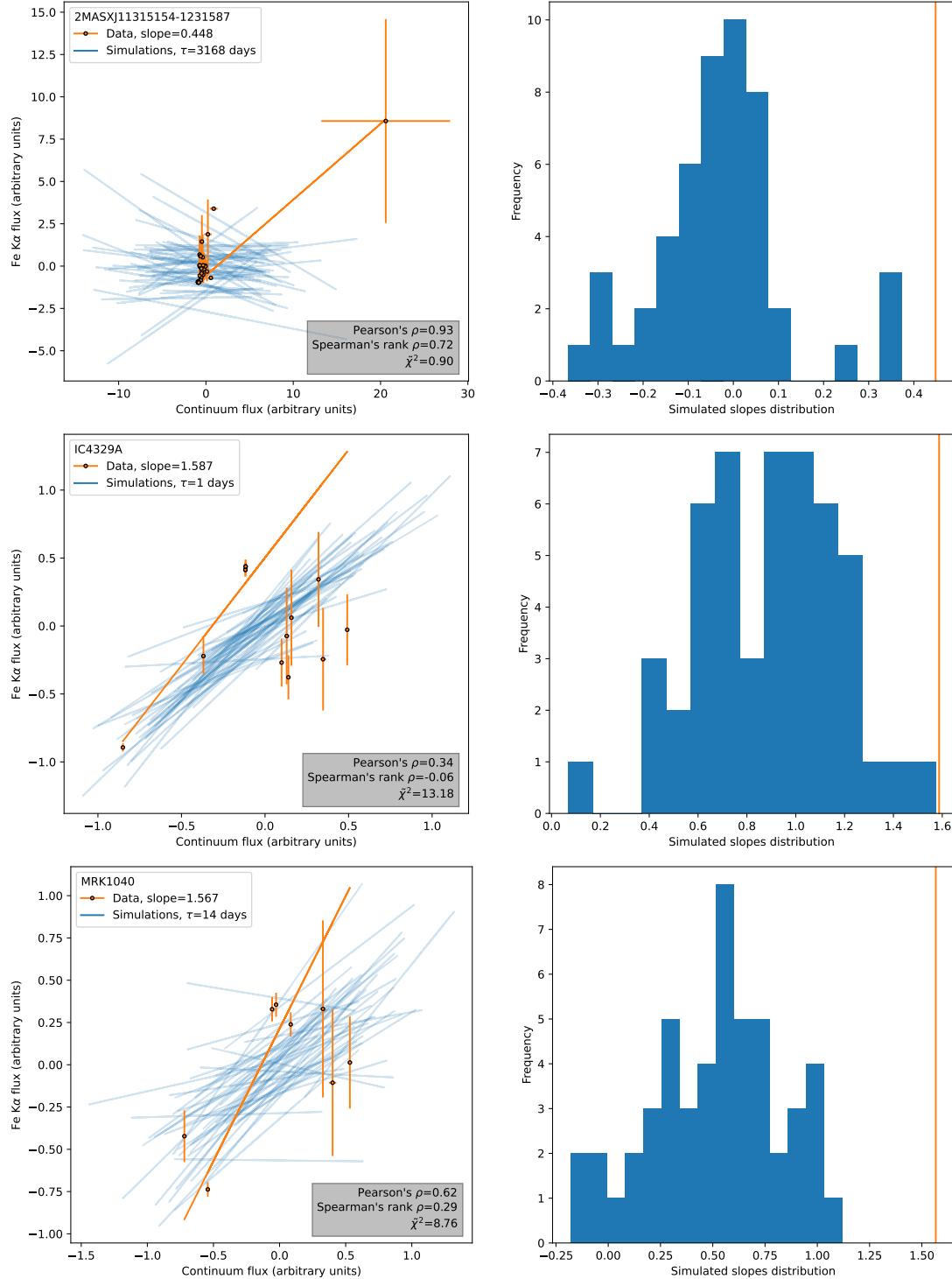


Figure 4.7: Flux-flux correlation at best τ

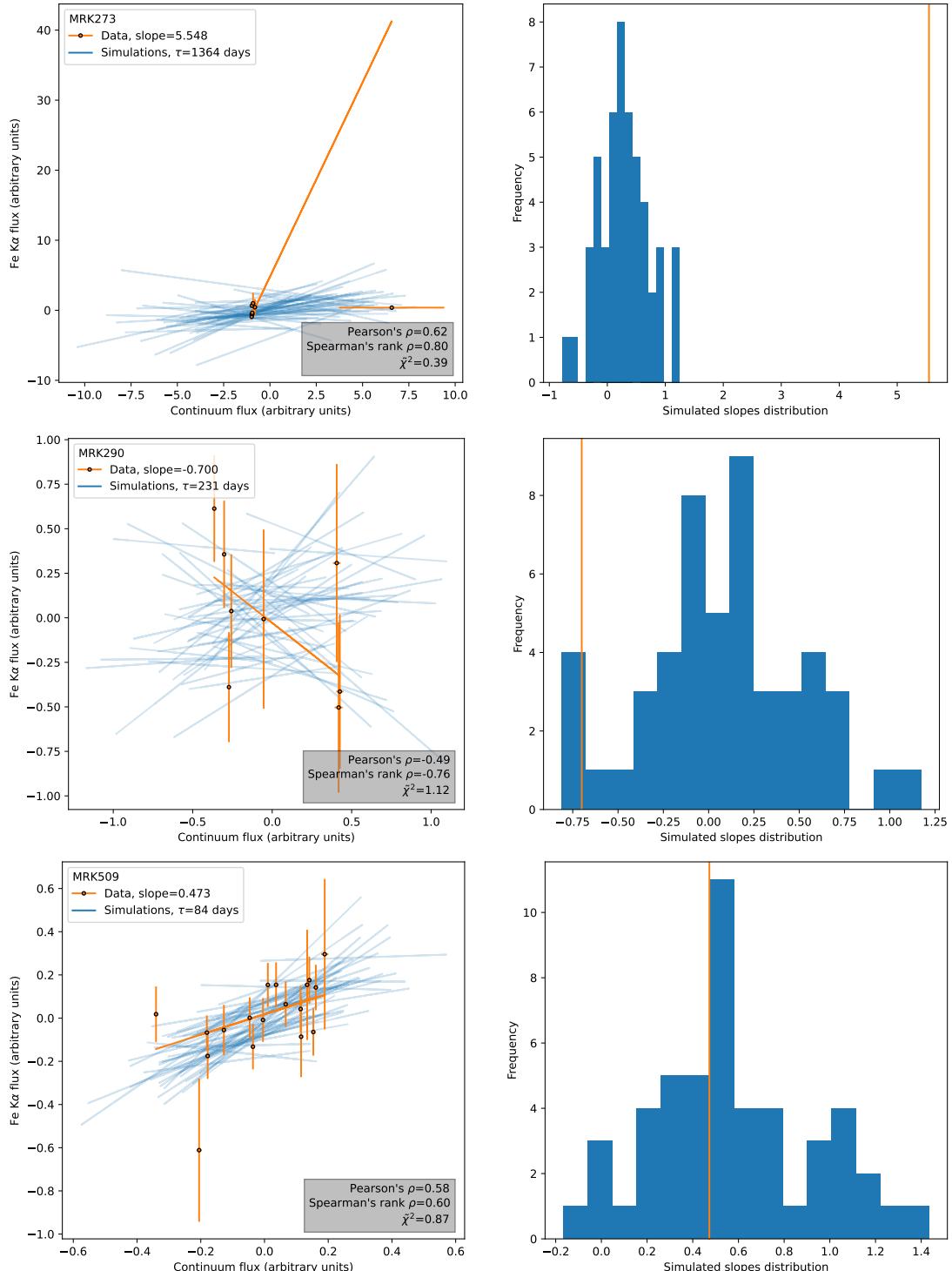


Figure 4.8: Flux-flux correlation at best τ

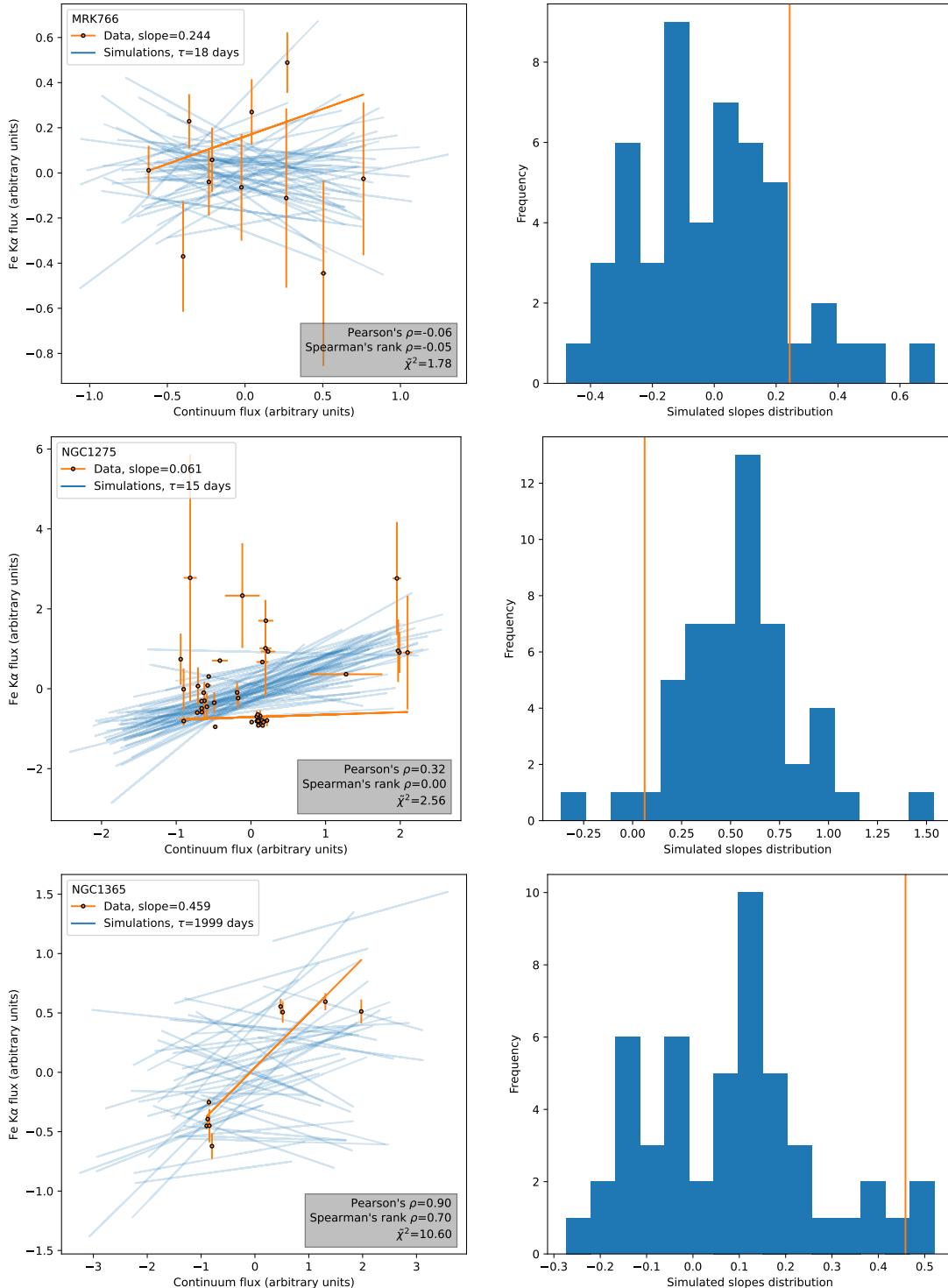


Figure 4.9: Flux-flux correlation at best tau

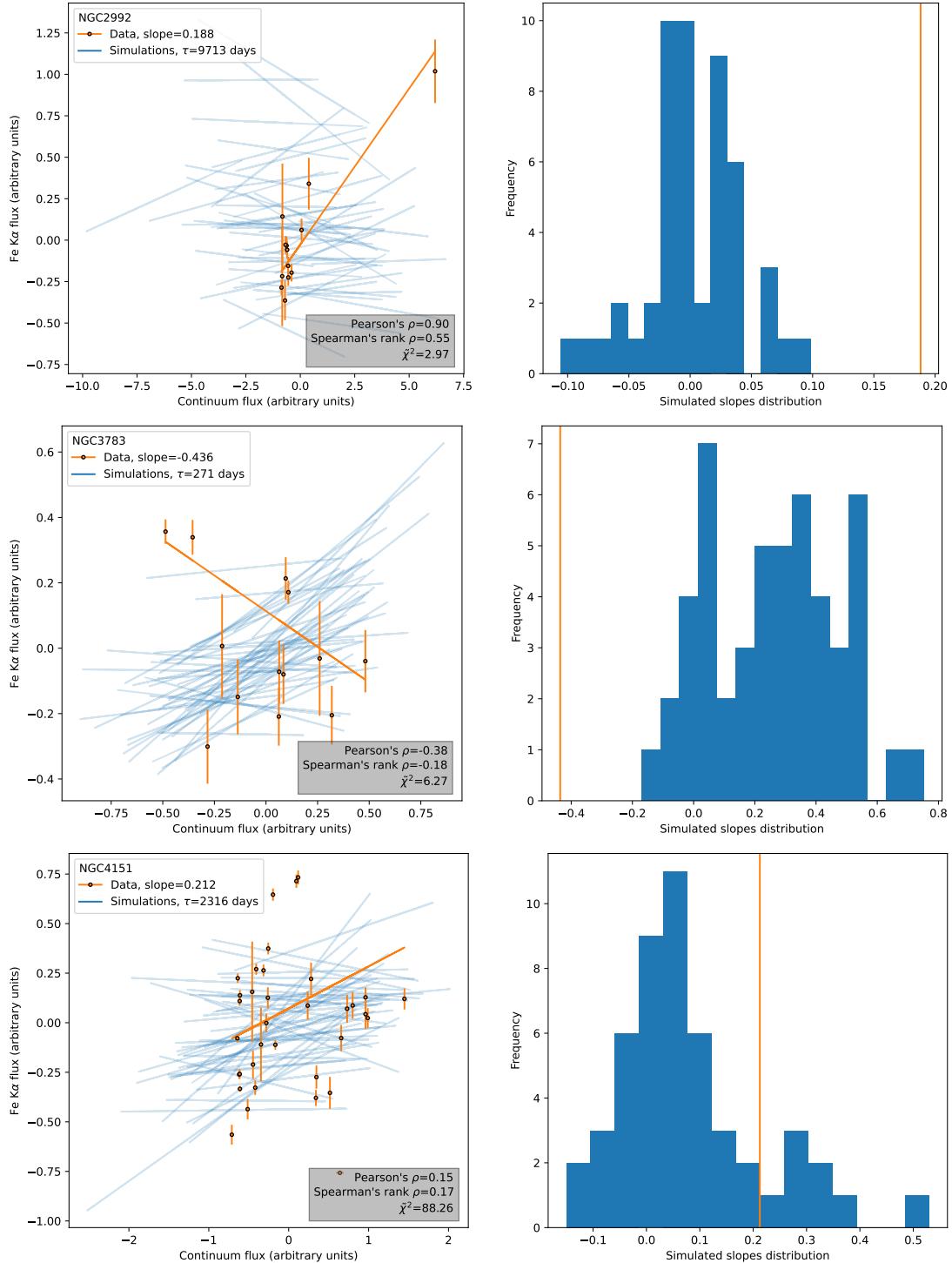


Figure 4.10: Flux-flux correlation at best tau

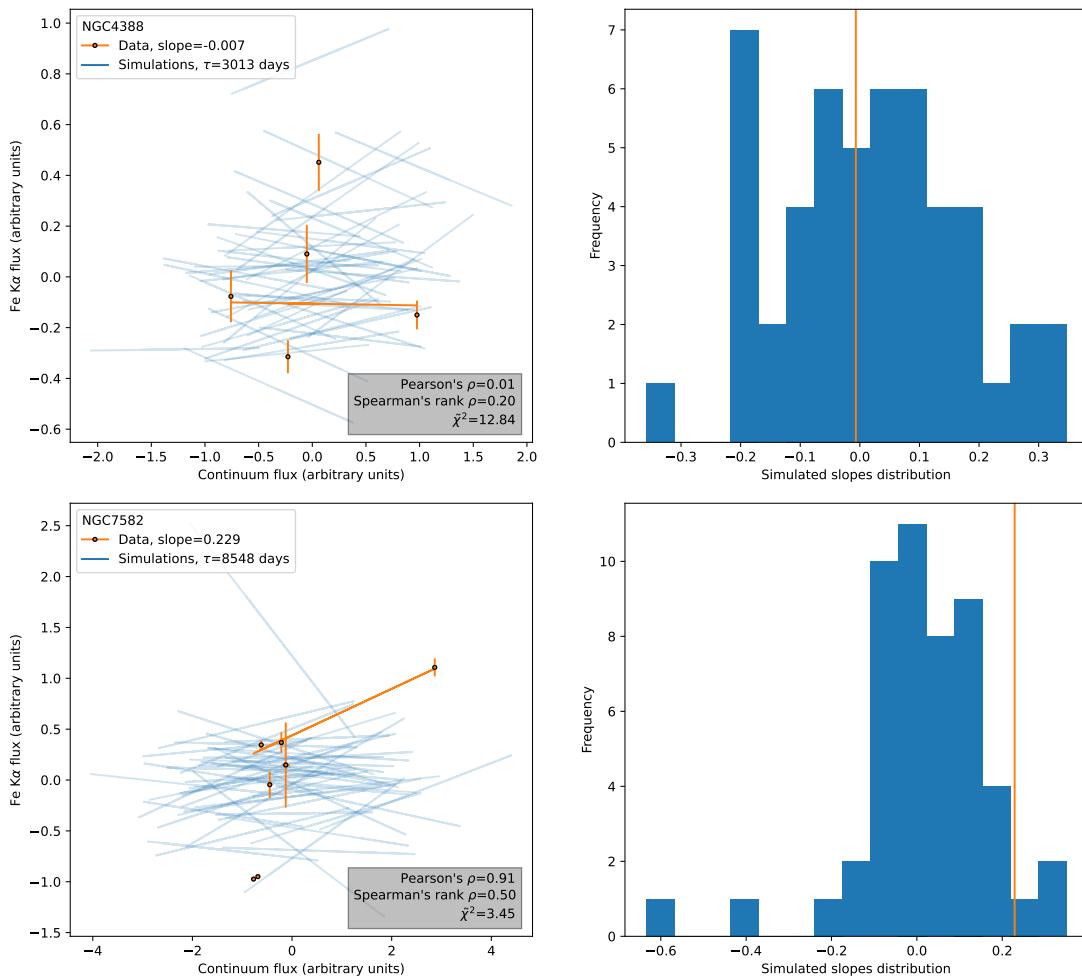


Figure 4.11: Flux-flux correlation at best tau

peating this analysis with an adequate amount of data-points would certainly allow for a better estimate of the reprocessor size.

Another possibility for future work is an analysis of the correlation between AGN type properties and ξ , as other physical factors may be influencing the emissions and therefore the measured size of the reprocessor, thus causing the scatter in the $M_{\text{BH}} - \tau$ relation. As already mentioned, the light curves of radio loud sources may be affected from beamed X-ray emission leading to light curves that do not trace the same variations. Another behavior we might expect is from Compton-thick AGN, which should have reflection-dominated continua with little variation, and thus our simple X-ray spectral fitting approach would measure the flux of this relatively static component causing a $\xi < 1$. Another example yet is the case of changing look AGN, where covering factor changes may increase the continuum variability from the observers point of view, but not necessarily from the point of view of the whole reflector.

4.7 Conclusions

In this Chapter we have used simulations in order to study the distribution of X-ray reprocessing gas (*the reprocessor*) in the active nuclei of a sample of local host galaxies. In our simple model we assumed that the X-ray continuum emission comes from the AGN corona and is then reprocessed by a spherical thin shell of gas around it, which causes a delay and a damping of the variations of the continuum in a way that depends on the distance between the corona and the reflector. In order to reproduce the behavior of this setup we simulated the X-ray light curves from the continuum starting from red noise in Fourier space following a power spectrum based on physical parameters of our sources, and the reflected component was obtained from the same light curve, which was smoothed and delayed in time.

The sample in this study was not observed through specific monitoring campaigns aimed at this kind of study, but we used archival data which allowed to retrieve light curves with few data-points observed through several years with irregular sampling. This did not allow to estimate a lag between the light curves but only to determine the variability reduction in amplitude caused by the reprocessor on the reflected component.

This simple modeling allowed us to completely constrain the size of the reprocessor for 4 sources in our sample and to determine at least a lower or upper limit size for 13 of them. For the remaining part of the sample, we either didn't have enough

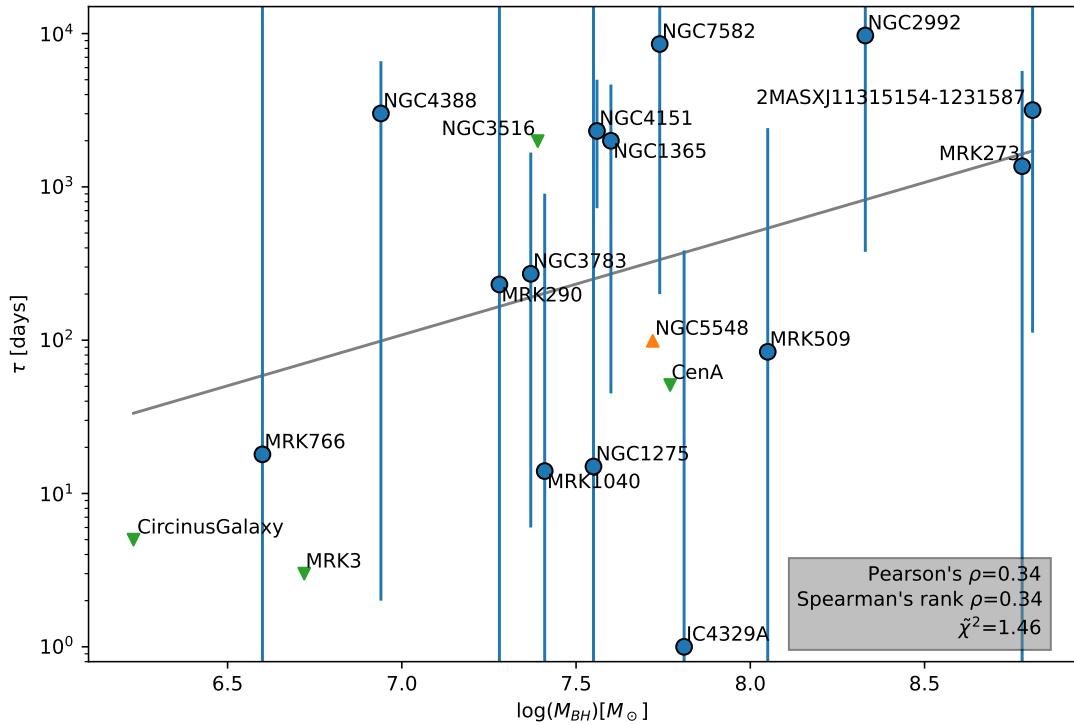


Figure 4.12: Correlation between black hole mass (M_{BH}) and reprocessor size τ expressed in light days. The y axes is limited to the range of light days considered in our work, $1 - 10,000$, and sources with unconstrained upper/lower limits have error bars extending all the way to the extremes of the plot.

data points to correctly estimate the variability reduction, or the data quality was insufficient, or our model is too simple and does not apply to some individual source. Our resulting reprocessor estimates are still a promising starting point to design future monitoring campaigns that will be more effective at constraining the size of the reprocessor and may even be able to determine a lag between the light curves.

We found a correlation between the size of the reprocessor and the black hole mass M_{BH} in our sample, which is consistent with other AGN properties which scale with the mass of the black hole and accretion rate, which include the Schwarzschild radius r_s (Schwarzschild, 1916), the internal radius of the dusty torus (Koshida et al., 2014) and the BLR size (Yu et al., 2020). The relation we found still has large scatter, $\sim 1\text{dex}$, which can be in part due to the large uncertainties in our estimation due to few data-points and stochasticity of BH variability but is also probably due to other physical factors like more complex geometries and AGN type dependent imprints that affect our measurements. The variability in the Fe K α flux guarantees that the reflector is located in the nuclear surroundings with a size of up to 10^4 light days, (i.e. $\lesssim 10$ pc), since a more extended reflector made of very high column density gas clouds on galactic scales, would imply no visible variability would in the Fe K α line flux. The cases in which we were only able to define a lower limit on the reprocessor size may be cases of clouds on galactic scales contributing to the reflection and therefore suggesting a different physical arrangement of the gas.

CHAPTER 5

Conclusions

In this thesis we have studied how accretion of the central supermassive black hole evolves in galaxies in all their star formation life phases and through a wide range of cosmic epochs. It is particularly interesting to look for a connection between these two accretion phenomena as they are both fed by cold gas, but they take place on very different galactic scales and it is not clear how the gas can lose angular momentum and funnel into the central black hole, so the two phenomena could in principle be completely independent. We have taken two complementary approaches, we have performed a statistical study on an observational sample from a deep and wide survey, the COSMOS survey, with an exceptional wavelength coverage (Chapter 2), and we have adopted Semi-Empirical models to generate mock galaxy catalogs and thus better understand the physical parameters that shape the relations we found in the data (Chapter 3).

We have mainly focused on the average L_X emission of galaxies divided in bins of stellar mass, the $L_X - M_*$ relation, in three galaxy life phases: star forming, quiescent and starburst. We have seen that the $L_X - M_*$ relation, which easily translates into a BHAR– M_* relation, follows a decreasing trend in time, probably as less gas is available in galaxies at later cosmic epochs thus causing a decrease in the average Eddington ratio, and that similarly to SFR, starbursts have a higher L_X than star forming galaxies, and quiescents have a lower L_X . Our SEMs point in the direction that these evolutions in time and across galaxy type are driven by a change in the average

Eddington ratio ζ_c which seems to rule the $L_X - M_*$ normalization. Independent results from observations and models, and also our own from Sec. 2.7, suggest that the $M_{\text{BH}} - M_*$ relation is constant with redshift but with a steeper slope than the one that would return our observed L_X slope, therefore implying an average Eddington ratio ζ_c decreasing with increasing stellar mass.

We have seen that main-sequence and quiescent galaxies share similar ratios of BHAR and SFR at all probed cosmic epochs, suggesting that the two processes are linked together throughout different galaxy phases. This ratio indicates that at fixed M_{BH}/M_* , a similar BHAR/SFR ratio as the one observed in star forming and quiescent galaxies, would be induced by a proportional decline in characteristic Eddington ratio ζ_c and specific SFR within a bin of stellar mass. Analogously, the significantly lower BHAR/SFR in starbursts with respect to quiescent/star forming galaxies would be naturally interpreted as a proportionally higher specific SFR and roughly constant or slightly higher ζ_c in these young, gas rich systems.

Our results on sBHAR and sSFR show similar evolutions and signs of downsizing in all galaxy types and at all explored redshifts, where we define downsizing as the fact that more massive galaxies accreted most of their black hole mass and of the stellar mass at very early cosmic epochs and their accretion decreased fast and steeply, whereas low-mass galaxies have accreted their mass more slowly, but their accretion rate decreased more slowly with time. Downsizing can also be seen in our mock catalogs from SEMs, that suggest that more massive galaxies have lower characteristic Eddington ratios ζ_c , and is mostly apparent in quiescent galaxies.

Our work points in the direction of a co-evolution of star formation and black hole accretion in galaxies across cosmic time where the bulk of the black hole and stellar masses is accreted in galaxies during the main-sequence phase through secular processes. Both accretions follow similar evolutionary patterns which appear to be driven by cold gas availability which affects star formation and the characteristic Eddington ratio ζ_c , even though the magnitude of this effect is different in both accretions: the starburst phase seems to be accompanied by a substantial enhancement of the SFR but a limited enhancement of the BHAR, especially at high redshifts. On the other hand, the quiescent phase seems characterized by very little star formation but still a significant residual black hole accretion. The final evolutionary picture that emerges from this work is one in which the host galaxy starts off from a main-sequence or even starburst, gas-rich phase, evolving at an almost constant (specific) SFR and then gradually switches off its black hole accretion and star formation due to internal gas

consumption, thus gradually reducing the SFR and BHAR.

On a distinct approach, in Chapter 4 we have studied the gas distribution around the AGN by simulating the variability of X-ray emissions from the corona and from the reprocessing gas in a sample of local galaxies. We used a simple model which assumed the reprocessing gas to be arranged in a spherical thin shell around the active galactic nucleus, which acts by delaying and smoothing out the variations of the continuum emission from the corona. This very simple model, together with sparse light curves from archival data allowed to place constraints for 17 out of 33 sources, constraints which could be useful in order to design future monitoring surveys. We also found a correlation with a large scatter between the size of the reprocessing gas and the mass of the black hole, suggesting that the mass of the BH plays a role in defining the gas arrangement around it, but also that our model may be too simple for some of our galaxies, and that the size of the reprocessor could depend on other physical factors like more complex geometries and/or AGN type dependent emission imprints.

APPENDIX A

Chapter 2 additional data

A.1 Sample properties

We report the Tables with additional properties of our galaxy sample from Chapter 2. Table A.1 lists star-forming galaxies, Table A.2 quiescent galaxies, and Table A.3 starbursts.

Table A.1: Properties of the star-forming galaxy sample. For each mass and redshift bin we show the median mass and redshift, and the number of stacked and detected galaxies in the 2-7 keV band. We also show the X-ray luminosity, the SFR estimated from the FIR, and the obscured SFR from the UV. Quantities are medians, and confidence ranges are 1σ .

Mass range $\log_{10}(M_*/M_\odot)$	M_* $\log_{10}(M_*/M_\odot)$	z	N_{stacked}	N_{detected}	$L_{\text{2-10 keV}}$ $10^{42} \text{ erg s}^{-1}$	SFR_{IR} $M_\odot \text{ yr}^{-1}$	SFR_{UV} $M_\odot \text{ yr}^{-1}$
$0.10 < z < 0.65$							
9.5–10.0	9.72	0.46	4244	19	$0.023^{+0.007}_{-0.007}$	$2.22^{+0.08}_{-0.07}$	$0.500^{+0.012}_{-0.011}$
10.0–10.5	10.21	0.45	2514	75	$0.142^{+0.009}_{-0.009}$	$5.68^{+2.67}_{-1.17}$	$0.479^{+0.013}_{-0.018}$
10.5–11.0	10.70	0.45	1214	117	$0.639^{+0.015}_{-0.015}$	$14.2^{+0.7}_{-0.7}$	$0.56^{+0.03}_{-0.02}$
11.0–12.0	11.14	0.43	239	38	$1.550^{+0.038}_{-0.036}$	$15.7^{+2.2}_{-2.4}$	$0.76^{+0.07}_{-0.11}$
$0.65 < z < 1.30$							
9.5–10.0	9.72	0.97	15431	56	$0.15^{+0.02}_{-0.02}$	$4.70^{+0.14}_{-0.13}$	$1.051^{+0.009}_{-0.011}$
10.0–10.5	10.22	0.97	7957	175	$0.66^{+0.03}_{-0.03}$	$14.9^{+0.3}_{-0.3}$	$0.854^{+0.018}_{-0.015}$
10.5–11.0	10.69	0.95	3888	370	$2.67^{+0.05}_{-0.05}$	$26.8^{+0.5}_{-0.5}$	$0.856^{+0.023}_{-0.016}$
11.0–12.0	11.15	0.95	761	142	$5.13^{+0.12}_{-0.12}$	$37.1^{+2.7}_{-2.1}$	$1.29^{+0.08}_{-0.13}$
$1.30 < z < 2.25$							
9.5–10.0	9.80	1.64	13451	47	$0.35^{+0.09}_{-0.08}$	$10.4^{+0.6}_{-0.6}$	$2.17^{+0.02}_{-0.02}$
10.0–10.5	10.22	1.73	11334	177	$1.47^{+0.09}_{-0.10}$	$30.2^{+0.7}_{-0.7}$	$1.62^{+0.03}_{-0.03}$
10.5–11.0	10.69	1.74	5273	368	$5.97^{+0.17}_{-0.16}$	$57.0^{+1.5}_{-1.5}$	$1.18^{+0.02}_{-0.02}$
11.0–12.0	11.12	1.76	954	172	$18.21^{+0.43}_{-0.45}$	$97.0^{+4.6}_{-4.7}$	$1.46^{+0.04}_{-0.06}$
$2.25 < z < 3.50$							
9.5–10.0	9.92	2.69	1264	3	$0.8^{+0.8}_{-0.9}$	$14.9^{+8.6}_{-6.7}$	$4.72^{+0.15}_{-0.17}$
10.0–10.5	10.22	2.77	7159	98	$2.9^{+0.4}_{-0.4}$	$68.2^{+4.4}_{-4.1}$	$4.49^{+0.07}_{-0.08}$
10.5–11.0	10.68	2.73	2726	166	$14.9^{+0.6}_{-0.6}$	$139.4^{+9.1}_{-9.1}$	$2.78^{+0.06}_{-0.05}$
11.0–12.0	11.13	2.69	466	70	$35.5^{+1.6}_{-1.5}$	$243.4^{+16.3}_{-14.8}$	$2.44^{+0.12}_{-0.14}$

Table A.2: Same as Table A.1 for the quiescent galaxy sample.

Mass range $\log_{10}(M_*/M_\odot)$	M_* $\log_{10}(M_*/M_\odot)$	z	N_{stacked}	N_{detected}	$L_{\text{2-10 keV}}$ $10^{42} \text{ erg s}^{-1}$	SFR_{IR} $M_\odot \text{ yr}^{-1}$	SFR_{UV} $M_\odot \text{ yr}^{-1}$
$0.10 < z < 0.65$							
10.0–10.5	10.27	0.47	867	12	$0.060^{+0.017}_{-0.018}$	$0.40^{+0.07}_{-0.09}$	$0.017^{+0.002}_{-0.002}$
10.5–11.0	10.74	0.47	830	19	$0.061^{+0.018}_{-0.015}$	$0.51^{+0.12}_{-0.10}$	$0.026^{+0.004}_{-0.004}$
11.0–12.0	11.16	0.50	271	14	$0.177^{+0.035}_{-0.035}$	$0.96^{+0.21}_{-0.22}$	$0.074^{+0.010}_{-0.017}$
$0.65 < z < 1.30$							
10.0–10.5	10.30	0.93	2594	19	$0.18^{+0.04}_{-0.05}$	$1.3^{+0.3}_{-0.4}$	$0.083^{+0.002}_{-0.002}$
10.5–11.0	10.75	0.94	3354	72	$0.41^{+0.04}_{-0.04}$	$2.2^{+0.3}_{-0.2}$	$0.130^{+0.003}_{-0.004}$
11.0–12.0	11.15	0.93	1107	25	$0.47^{+0.08}_{-0.08}$	$2.8^{+0.5}_{-0.4}$	$0.172^{+0.009}_{-0.010}$
$1.30 < z < 2.25$							
10.0–10.5	10.25	1.58	769	9	$0.6^{+0.3}_{-0.3}$	—	$0.279^{+0.013}_{-0.011}$
10.5–11.0	10.75	1.57	1534	25	$1.2^{+0.2}_{-0.2}$	$7.4^{+0.8}_{-0.8}$	$0.307^{+0.008}_{-0.008}$
11.0–12.0	11.14	1.59	532	8	$1.8^{+0.4}_{-0.4}$	$6.0^{+2.1}_{-3.0}$	$0.400^{+0.021}_{-0.012}$
$2.25 < z < 3.50$							
10.0–10.5	10.32	2.56	105	3	$8.4^{+2.5}_{-2.6}$	—	$1.25^{+0.22}_{-0.06}$
10.5–11.0	10.75	2.52	272	7	$2.6^{+1.5}_{-1.5}$	—	$1.15^{+0.06}_{-0.07}$
11.0–12.0	11.12	2.48	66	6	$22.2^{+3.6}_{-3.6}$	—	$1.71^{+0.18}_{-0.15}$

Table A.3: Same as Table A.1 for the starburst galaxy sample. We only use SFR_{IR} for this sample.

Mass range $\log_{10}(\text{M}_*/\text{M}_\odot)$	M_* $\log_{10}(\text{M}_*/\text{M}_\odot)$	z	N_{stacked}	N_{detected}	$L_{2-10 \text{ keV}}$ $10^{42} \text{ erg s}^{-1}$	SFR_{IR} $\text{M}_\odot \text{ yr}^{-1}$
$0.10 < z < 0.65$						
9.50–10.25	9.87	0.53	58	4	$0.36^{+0.09}_{-0.09}$	22^{+4}_{-2}
10.25–10.75	10.52	0.53	6	5	$4.68^{+0.40}_{-0.41}$	74^{+20}_{-9}
10.75–11.50	10.96	0.52	2	3	$6.64^{+0.67}_{-0.68}$	143^{+5}_{-1}
$0.65 < z < 1.30$						
9.50–10.25	9.90	0.95	162	7	$0.8^{+0.2}_{-0.2}$	66^{+3}_{-3}
10.25–10.75	10.42	1.04	161	10	$2.5^{+0.3}_{-0.3}$	131^{+5}_{-2}
10.75–11.50	10.91	1.08	43	5	$6.2^{+0.6}_{-0.6}$	280^{+15}_{-16}
$1.30 < z < 2.25$						
9.50–10.25	10.01	1.76	57	1	—	222^{+15}_{-18}
10.25–10.75	10.50	1.74	145	24	$15.1^{+1.2}_{-1.2}$	331^{+16}_{-9}
10.75–11.50	10.89	1.79	78	23	$40.8^{+2.0}_{-2.0}$	536^{+34}_{-23}
$2.25 < z < 3.50$						
9.50–10.25	10.07	2.48	5	0	—	384^{+14}_{-41}
10.25–10.75	10.53	2.59	64	2	$8.3^{+4.5}_{-4.1}$	739^{+27}_{-26}
10.75–11.50	10.90	2.68	46	7	$20.1^{+5.5}_{-5.3}$	1229^{+148}_{-139}

APPENDIX B

Chapter 4 additional data

B.1 $X(\tau)$ for all sources

In this Appendix we show for each source in the Andonie sample from Chapter 4 the detailed trends of X as a function of the delay/smoothing τ (Eq. 4.6) as determined by our simulations. We remind the reader that $X = 0$ is our best estimate of the reprocessor size, while $X = \pm 1$ are 1σ equivalent upper and lower limits (68.3% probability range).

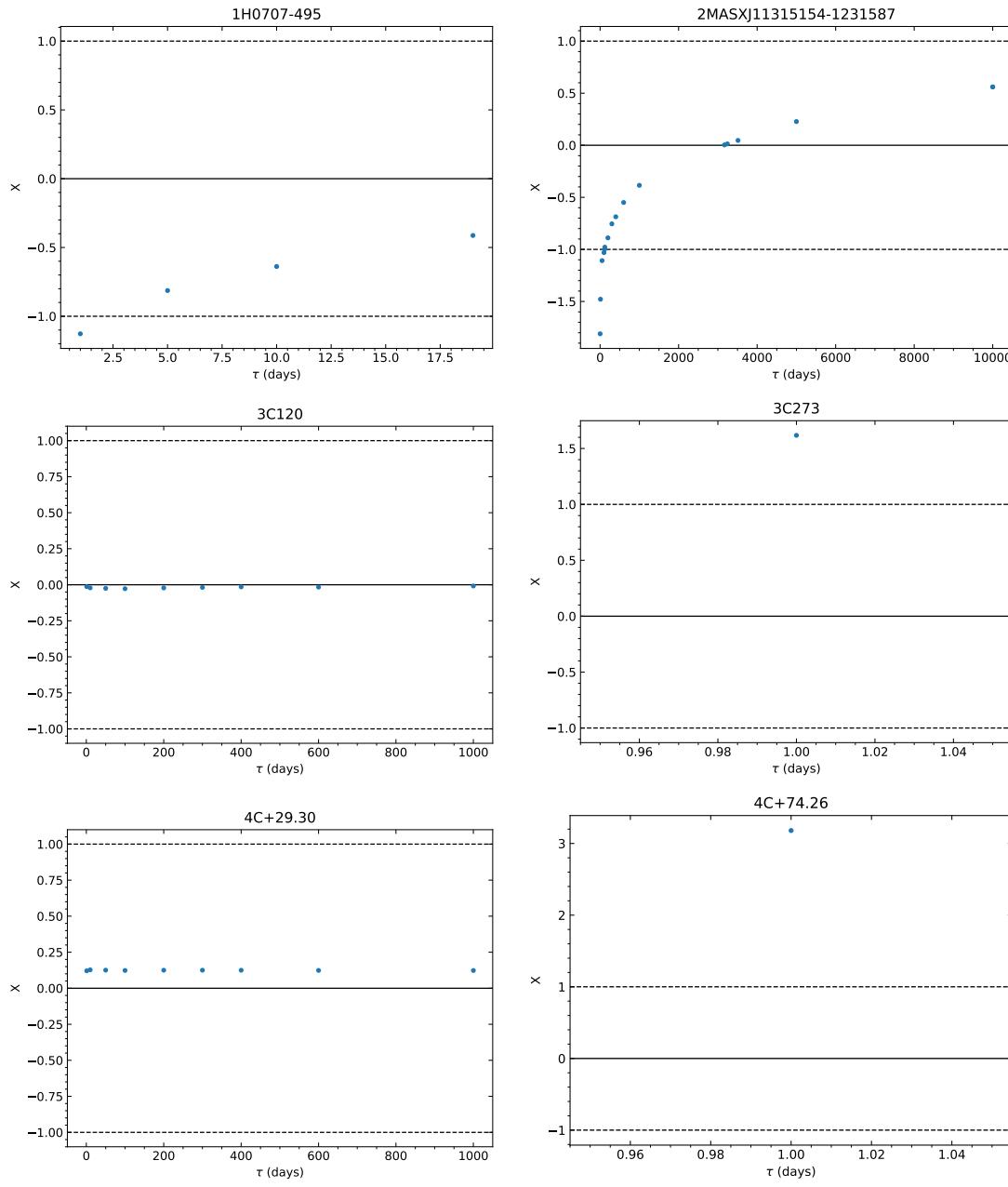


Figure B.1: $X(\tau)$ for the galaxy sample from Andonie et al. in prep. Best estimate of the reprocessor size ($X = 0$) is shown as a black solid line, while upper and lower limits ($X = \pm 1$) are shown as dashed lines

B.1. $X(\tau)$ FOR ALL SOURCES

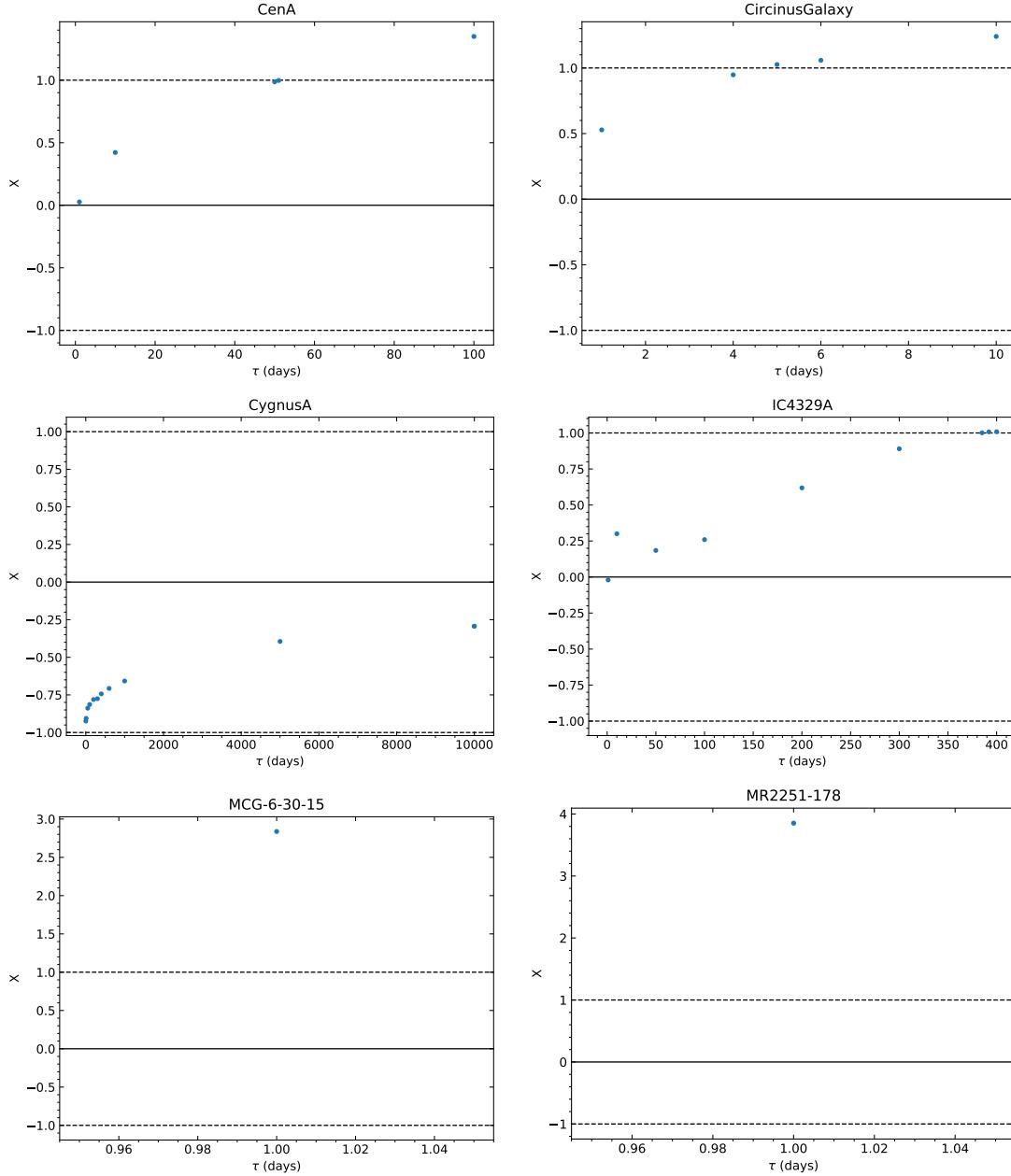


Figure B.2: Fig. B.1 continued.

APPENDIX B. CHAPTER ?? ADDITIONAL DATA

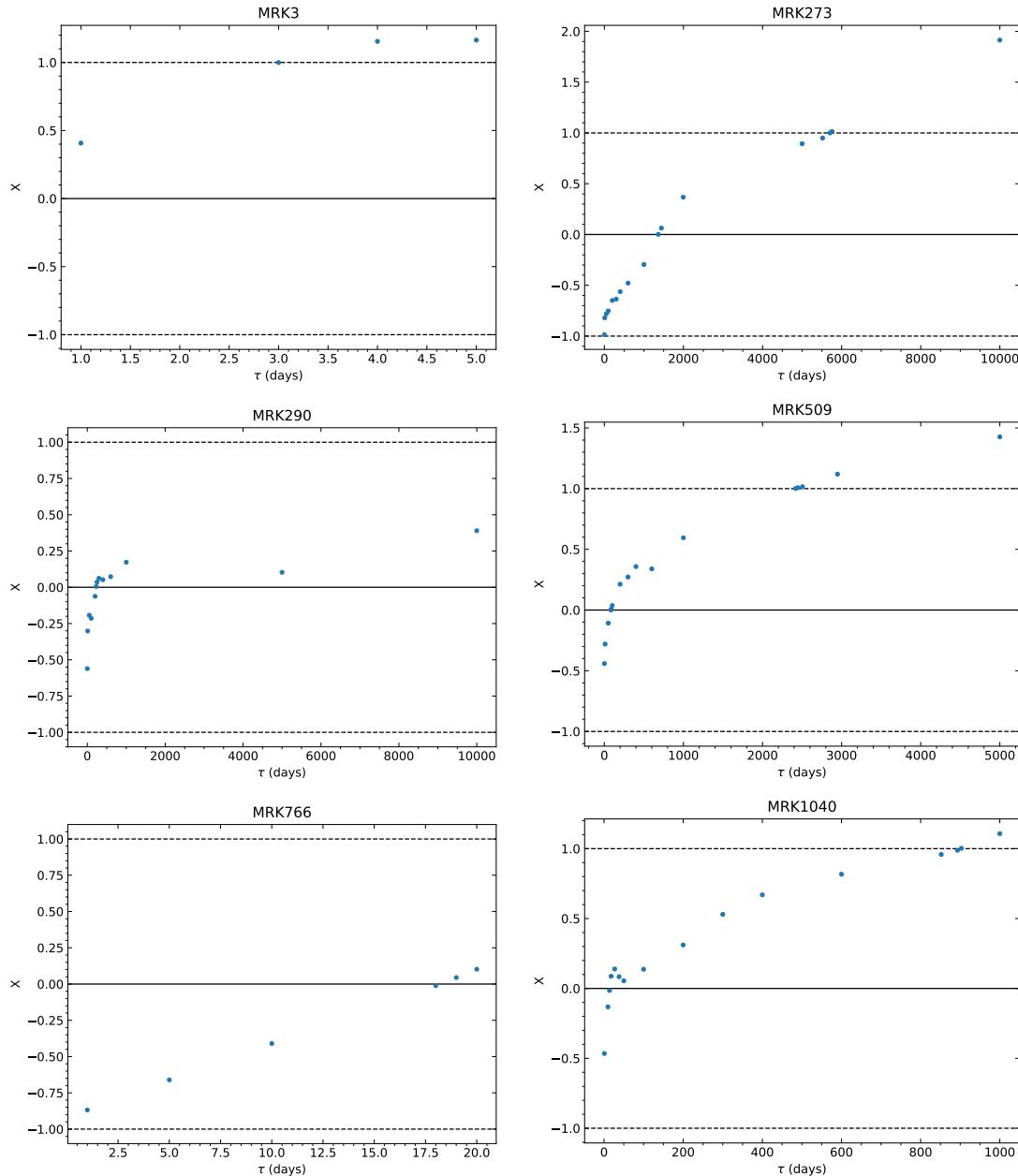


Figure B.3: Fig. B.1 continued.

B.1. $X(\tau)$ FOR ALL SOURCES

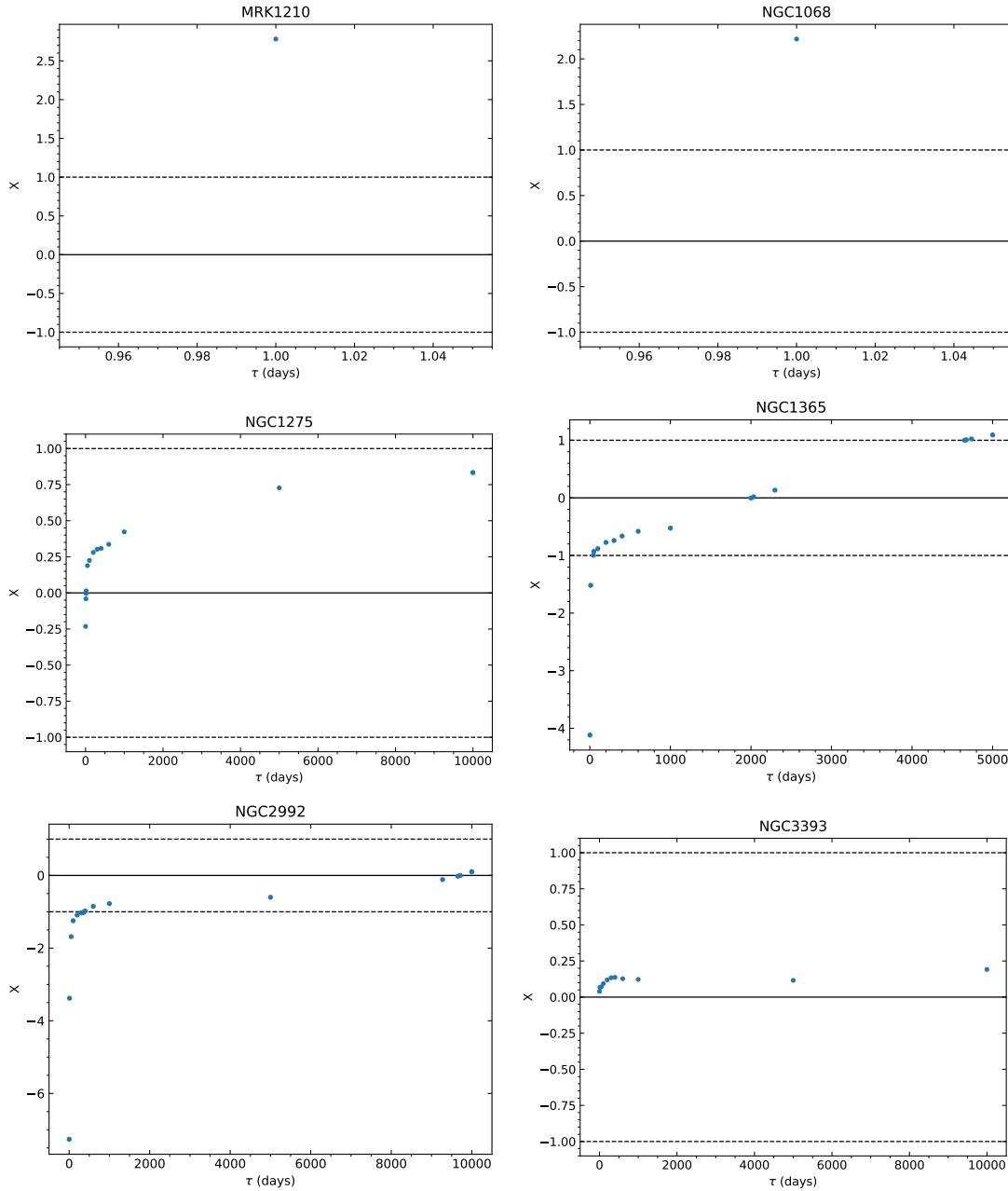


Figure B.4: Fig. B.1 continued.

APPENDIX B. CHAPTER ?? ADDITIONAL DATA

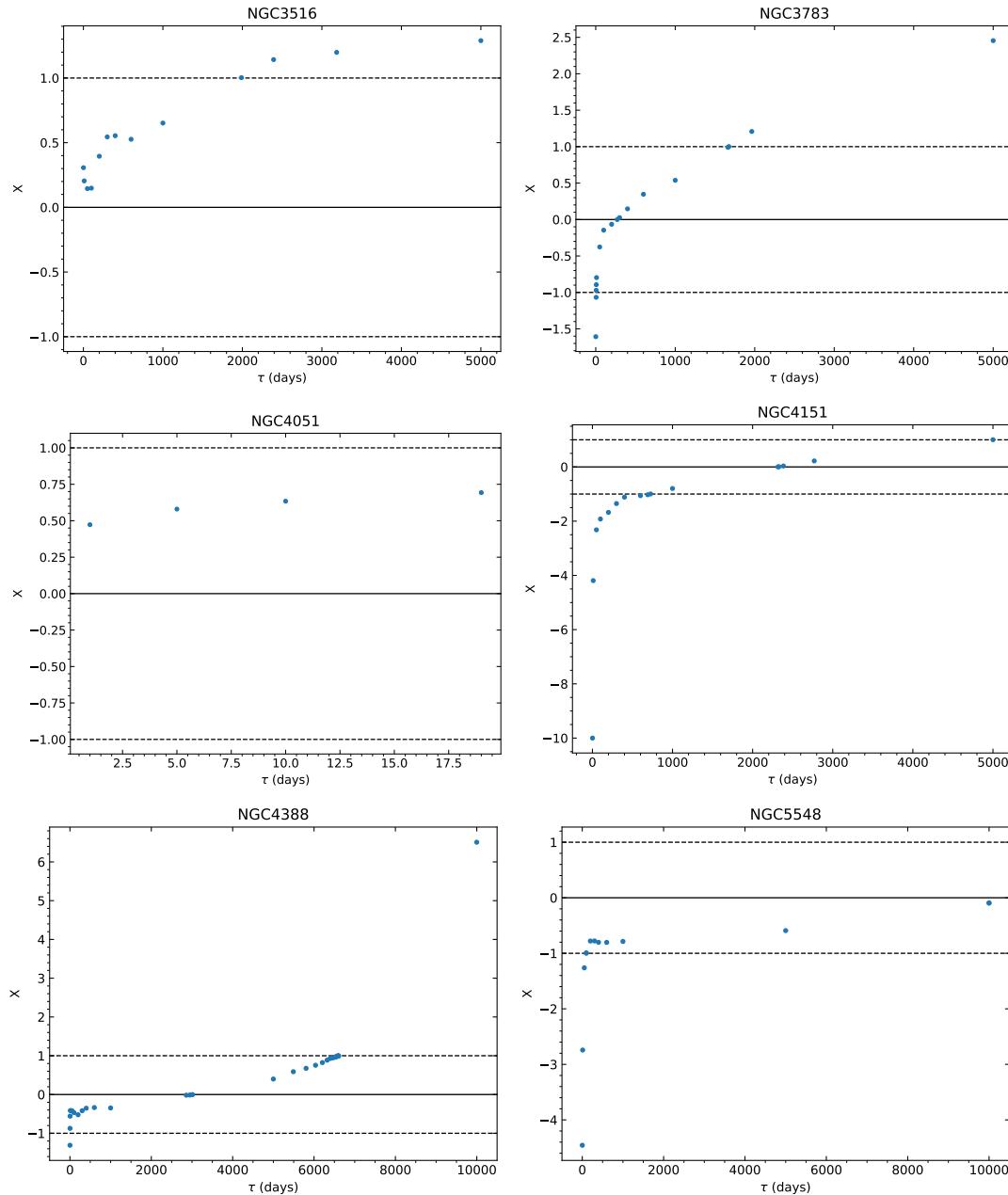


Figure B.5: Fig. B.1 continued.

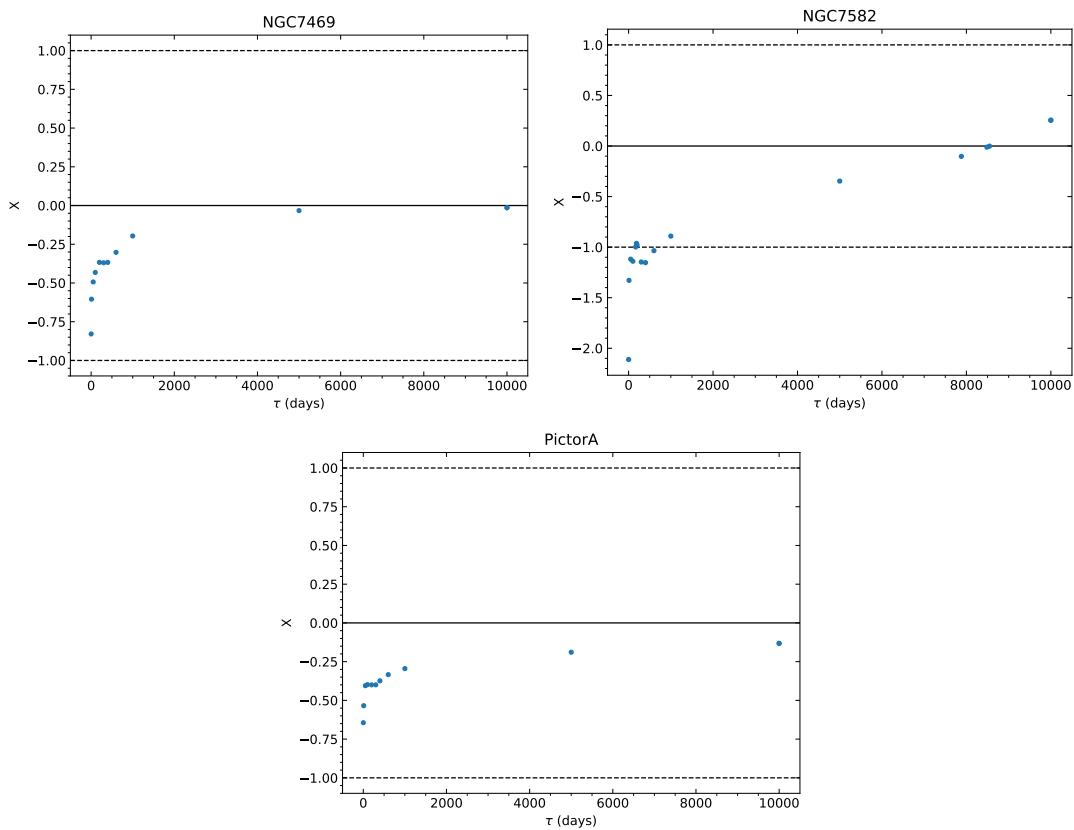


Figure B.6: Fig. B.1 continued.

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