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Quasar spectral variability from the *XMM-Newton* serendipitous source catalogue

R. Serafinelli^{1,2}, F. Vagnetti^{1,}, and R. Middei³

- Dipartimento di Fisica, Università di Roma "Tor Vergata", via della Ricerca Scientifica 1, 00133 Roma, Italy e-mail: roberto.serafinelli@roma2.infn.it
- ² Dipartimento di Fisica, Università di Roma "La Sapienza", Piazzale Aldo Moro 5, 00185 Roma, Italy
- Dipartimento di Matematica e Fisica, Università Roma Tre, via della Vasca Navale 84, 00146 Roma, Italy

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ABSTRACT

Context. X-ray spectral variability analyses of active galactic nuclei (AGN) with moderate luminosities and redshifts typically show a "softer when brighter" behaviour. Such a trend has rarely been investigated for high-luminosity AGNs ($L_{bol} \gtrsim 10^{44}$ erg/s), nor for a wider redshift range (e.g. $0 \lesssim z \lesssim 5$).

Aims. We present an analysis of spectral variability based on a large sample of 2700 quasars, measured at several different epochs, extracted from the fifth release of the XMM-Newton Serendipitous Source Catalogue.

Methods. We quantified the spectral variability through the parameter β defined as the ratio between the change in the photon index Γ and the corresponding logarithmic flux variation, $\beta = -\Delta\Gamma/\Delta\log F_{\rm X}$.

Results. Our analysis confirms a softer when brighter behaviour for our sample, extending the previously found general trend to high luminosity and redshift. We estimate an ensemble value of the spectral variability parameter $\beta = -0.69 \pm 0.03$. We do not find dependence of β on redshift, X-ray luminosity, black hole mass or Eddington ratio. A subsample of radio-loud sources shows a smaller spectral variability parameter. There is also some change with the X-ray flux, with smaller β (in absolute value) for brighter sources. We also find significant correlations for a small number of individual sources, indicating more negative values for some sources.

Key words. surveys – galaxies: active – quasars: general – X-rays: galaxies

1. Introduction

All classes of active galactic nuclei (AGN) show some kind of amplitude variability, which is present in nearly all electromagnetic bands, from radio to X-rays and on several timescales. In particular, X-ray variability is observed both on very short timescales (Ponti et al. 2012), allowing the study of the innermost regions of the AGN, and on longer timescales of at least a number of years (e.g. Markowitz & Edelson 2004; Vagnetti et al. 2011, 2016; Shemmer et al. 2014).

A geometrically thin, optically thick accretion disk around the central supermassive black hole is believed to emit optical and ultraviolet photons, while a hot corona would produce the X-rays by means of a comptonisation process of the radiation emitted by the disk itself (Haardt & Maraschi 1991). While this process is not entirely understood, particularly because the shape, size and geometry of the corona are mostly unknown, variability could provide a means of procuring information about it.

Besides amplitude variability, the study of spectral variability is also interesting. This feature has been studied by several authors in the optical/UV bands (e.g. Giveon et al. 1999; Trevese & Vagnetti 2002; Vagnetti et al. 2003) resulting in the discovery of a "harder when brighter" behaviour for the power-law spectrum, defined as $F_{\nu} \propto \nu^{\alpha}$.

In the X-ray band, spectral variability has mostly been studied for individual sources (e.g. Magdziarz et al. 1998; McHardy 2001; Zdziarski et al. 2003; Ursini et al. 2016), while

few ensemble studies have been made on the topic, such as the one by Sobolewska & Papadakis (2009). The trend these studies have found is quite the opposite of the optical one, since a softer when brighter behaviour was found in both individual and ensemble studies, even at higher energies (Soldi et al. 2014). One notable exception is the work by Connolly et al. (2016), in which the authors analysed a sample of local low-luminosity AGNs and found harder when brighter behaviour for low-Eddington-ratio sources and softer when brighter behaviour for high-Eddington-ratio sources. These observations explain the observed difference in terms of dominant seed photons source, which in the first case is cyclo-synchroton emission from the Comptonising corona itself, while in the second case is thermal emission from the accretion disk.

These studies, though, are mainly focused on nearby, very bright objects such as Seyfert galaxies, and very few studies were carried out concerning quasars and type-1 AGNs in a very large luminosity and redshift range. Some of these studies include the ones performed by Paolillo et al. (2004) and Gibson & Brandt (2012), as well as the ones we recently published, Vagnetti et al. (2016) and Serafinelli et al. (2016). In all these works the authors found evidence of spectral variability for some of the sources, with a softer when brighter trend.

Section 2 briefly summarises the characteristics of our catalogue, Sect. 3 focuses on the detailed description of our analysis and Sect. 4 summarises and discusses our result.

Throughout the paper, we adopted the following cosmology: $H_0 = 70 \text{ km s}^{-1} \text{ Mpc}^{-1}$, $\Omega_m = 0.3$ and $\Omega_{\Lambda} = 0.7$.

Table 1. Number of epochs and sources listed for each catalogue.

Catalogue	Observations	Unique sources
SDSS-DR7Q	/	105 783
SDSS-DR12Q	/	297 301
3XMM-DR5	565 962	396 910
MEXSAS	7837	2700

2. Data

The present work makes use of the third *XMM-Newton* serendipitous source catalogue data release 5 (3XMM-DR5), in which a total of 565 962 XMM observations of 396 910 unique X-ray sources are listed (Rosen et al. 2016). This catalogue includes observations from the beginning of the mission in the year 2000 that were made public as of the 31st December, 2013. All kinds of astrophysical sources are present, both galactic and extragalactic, from AGN to X-ray binaries and isolated objects such as pulsars, magnetars and X-ray emitting accreting disks around stellar-mass black holes (Lin et al. 2012).

In a previous paper (Vagnetti et al. 2016), we introduced the Multi-Epoch XMM-Newton Serendipitous AGN Sample (MEXSAS), which was obtained by cross-matching the 3XMM-DR5 catalogue with two quasar catalogues of the Sloan Digital Sky Survey, that is, SDSS-DR7Q (Schneider et al. 2010) and SDSS-DR12Q (Pâris et al. 2017). These two optical catalogues list 105 783 and 297 301 quasars, respectively, according to slightly different definitions. The DR7Q defines a quasar as having an absolute *I*-band magnitude $M_I \leq -22$ (Schneider et al. 2010), while in the DR12Q, the absolute magnitude in the I-band is $M_I \leq -20.5$ (Pâris et al. 2017), therefore the latter catalogue also lists some low-z Seyfert galaxies. Moreover, there is also a subset of sources identified as broad absorption line (BAL) quasars, and some radio-loud AGNs are also present (see Sec. 3.4.4). The cosmology used by both Schneider et al. (2010) and Pâris et al. (2017) is identical to that used in this paper.

The MEXSAS catalogue has 7837 observations, corresponding to 2700 unique quasar sources. The MEXSAS is a multiepoch catalogue, which means that a minimum of two observations for each source is required. The maximum number of observations for a single source is 39. The typical number of EPIC counts for each observation is given by $\langle \log N \rangle \sim 2.2$. A summary of the number of observations and sources of the catalogues can be found in Table 1.

3. Spectral variability

To quantify the spectral variability, that is, the variations of flux with the frequency, we use the spectral variability parameter β , first introduced by Trevese & Vagnetti (2002):

$$\beta = \frac{\alpha(t + \Delta t) - \alpha(t)}{\log F_B(t + \Delta t) - \log F_B(t)} = \frac{\Delta \alpha}{\Delta \log F_B},\tag{1}$$

that relates the variations of the spectral index α , with the variations of the flux F_B , integrated in a given band B. In the X-rays, however, it is customary to express the energy spectrum as $N(E) \propto E^{-\Gamma}$, E = hv being the energy of the photon and $\Gamma = 1 - \alpha$ the photon index. Hence, we express Eq. (1) as:

$$\beta = -\frac{\Delta\Gamma}{\Delta\log F_{\rm X}},\tag{2}$$

with F_X being the flux in a given X-ray band.

The photon index Γ , though, is not available in our catalogue and in order to calculate Eq. (2) for each observation, we could either compute the spectral variability parameter by means of the hardness ratios, as described in the following section, or, alternatively, estimate the values of Γ from the catalogue flux data.

3.1. Ensemble analysis using hardness ratios

The hardness ratio between two given bands, H and S, is usually defined as:

$$HR = \frac{CR_H - CR_S}{CR_H + CR_S},\tag{3}$$

where CR_i is the count-rate in the band i.

The 3XMM-DR5 catalogue lists four hardness ratios: HR_1 is computed between the count-rates in bands 1 and 2, HR_2 between bands 2 and 3, HR_3 between bands 3 and 4 and finally HR_4 between bands 4 and 5. Band 1 includes all photons between 0.2 and 0.5 keV, while band 2 photons have energy between 0.5 and 1 keV, band 3 photons are in the range 1–2 keV, band 4 is the energy band 2–4.5 keV, and, finally, band 5 is the highest energy band, including all photons between 4.5 and 12 keV.

The hardness ratio can be related to β :

$$\frac{\Delta HR}{\Delta \log F} \simeq \frac{\mathrm{d}HR}{\mathrm{d}\Gamma} \frac{\Delta \Gamma}{\Delta \log F} = -\beta \frac{\mathrm{d}HR}{\mathrm{d}\Gamma},\tag{4}$$

where we have used Eq. (2). This means that the spectral variability parameter β can be written as:

$$\beta = -\left(\frac{\Delta HR}{\Delta \log F}\right) \left(\frac{\mathrm{d}HR}{\mathrm{d}\Gamma}\right)^{-1}.$$
 (5)

The denominator of Eq. (5) can be computed analytically as a function of Γ in the case of a spectral pure power-law model, or numerically, generating various spectral models using the XSPEC v.12.9.0 software package (Arnaud 1996) in cases where we want to consider a moderate absorption.

The count-rate in the band i can be related to the flux by the relation $CR_i = k_i F_i$, with the coefficients k_i that can be computed by the online tool WebPIMMS¹. Therefore, the numerator of Eq. (5) is approximately the linear fit between the deviations from the mean values of HR and $\log F$ for a given source. Details on this procedure can be found in Serafinelli et al. (2016), where various spectral models such as pure power-law and moderate absorption with different values of the column density $N_{\rm H}$ have been considered, and hints of a softer when brighter behaviour were found, computing a global β for different possible values of Γ . However, with this method, beta is derived indirectly from HR variations, and depends on Gamma through the derivative $dHR/d\Gamma$ in Eq. (5). It is for this reason that in this work we decided to compute the photon index Γ directly, as described in the following section.

3.2. Computing the photon index Γ

We now want to compute Γ as the slope of spectral fits performed on the available spectral data, assuming a power-law behaviour for the X-ray spectrum of each observation.

https://heasarc.gsfc.nasa.gov/cgi-bin/Tools/ w3pimmw/w3pimms.pl

The available fluxes F_X are not monochromatic and are related to the monochromatic flux F_i of the minimum energy of the band by the relation:

$$F_{X} = \int_{\nu_{i}}^{\nu_{s}} F_{i} \left(\frac{\nu}{\nu_{i}}\right)^{1-\Gamma} d\nu = \frac{F_{i}\nu_{i}}{2-\Gamma} \left[\left(\frac{\nu_{s}}{\nu_{i}}\right)^{2-\Gamma} - 1 \right], \tag{6}$$

where ν_i and ν_s are the minimum and maximum frequencies of the band, respectively. We have also assumed a power-law spectrum of the form $F_{\nu} = F_i(\nu/\nu_i)^{1-\Gamma}$ over the whole X-ray band.

If we invert Eq. (6) and convert frequencies in energy using E = hv, we obtain the expression for the monochromatic flux:

$$F_{i} = \frac{hF_{X}(2 - \Gamma)}{E_{i} \left[\left(\frac{E_{s}}{E_{i}} \right)^{2 - \Gamma} - 1 \right]},\tag{7}$$

where h is the Planck constant, which in useful units of measure can be expressed as $h = 4.14 \times 10^{-18} \text{ keV/Hz}$, if E_i is expressed in keV.

We used Eq. (7) assuming a typical photon index $\Gamma_0 = 1.7$, which best represents the bulk of the XMM sources, as reported by Rosen et al. (2016) and Mateos et al. (2009). Then we computed the linear fit between $\log F_i$ and $\log E_i$ to obtain the actual photon index Γ and its error $\sigma(\Gamma)$. In order to gain precision, we iterated using the photon indices obtained in the first step, instead of Γ_0 . A further iteration did not substantially modify the results, since the convergence was reached after just two iterations.

It should be stressed that, given the strong inhomogeneity in the flux errors from band to band and even from observation to observation, and also because we are dealing with data consisting of only a small number of spectral points, the linear fits must be weighted using

$$w_i = \frac{1}{\sigma^2(\log F_i)},$$

where $\sigma(\log F_i)$ is the error associated with $\log F_i$, in order to give less significance to points that deviate significantly from the power-law form but have a remarkable error.

Even though we are dealing with type-1 objects, a moderate absorption may still be present in the softer X-ray bands. However, we cannot take absorption into account since, in our analysis, we only use simple power-law spectral models. Therefore, we excluded the lowest-energy point from our spectral fits, for which the deviation from pure power-law spectrum, likely due to absorption, if present, is not negligible, and performed them with the remaining four points only. At relatively larger redshifts, this approximation works better, since the rest-frame photons are emitted at higher energies. At low redshifts, this estimation might, at most, slightly underestimate the photon indices in a few cases, however, we are mainly interested in temporal variations of Γ , thus any residual difference will be neglected.

In order to check the validity of this method, we confronted our Γ s with the XMMFITCAT spectral catalogue² by Corral et al. (2015), in which the authors computed the Γ s of a limited subset with a high number of counts ($N \geq 50$) of the 3XMM-DR5 catalogue. We created a subsample of coincident data with the XMMFITCAT catalogue, which consists of 2002 observations of 707 sources, and we compared our Γ s in this subsample with the WAPO_GAMMA entry, corresponding to a power-law with intrinsic absorption spectral fit of the XMMFITCAT

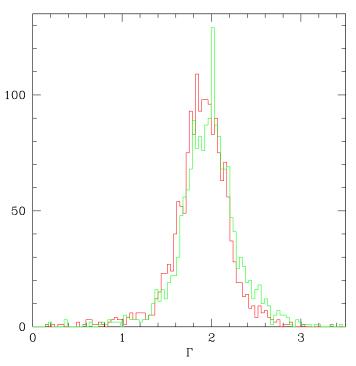


Fig. 1. Green histogram: distribution of WAPO_GAMMA in the coincident data between our catalogue and the XMMFITCAT catalogue. Red histogram: distribution of Γ computed for the same subsample with our method.

catalogue (for this matched subsample, the number of counts is $N \gtrsim 100$). Not all the observations are well fitted by an absorbed power-law spectrum, and Corral et al. (2015) flag those observations that are well fitted by a particular spectral model, requiring the XSPEC command goodness to return a value lower than 50%. Since 90% of the observations satisfy this requirement, we only compare our Γ with the ones from this subset, which consists of 1771 observations of 630 unique sources. As can be seen in Fig. 1, the two Γ distributions are very similar (with a small shift between the mean values $\langle \text{WAPO_GAMMA} \rangle - \langle \Gamma \rangle \simeq 0.07$), suggesting that our method is robust, even if the Γ s are obtained from only four points instead of from a whole spectrum.

We stress that our pure power-law approximation could not properly represent the actual spectral shape, in some cases affected by the presence of additional components, such as soft excess (e.g. Turner & Pounds 1988) or complex absorption, such as the warm absorber (e.g. Reynolds & Fabian 1995), for example. The incidence of these cases is likely limited to approximately 10% of the observations, as in the XMMFITCAT subsample. Moreover, we assume that the effect of these phenomena is statistically negligible for the estimate of the Γ variations, which are mainly involved in our subsequent analysis (see Sect. 3.3).

3.3. Ensemble analysis using the computed photon indices

Once the photon indices have been calculated for all the observations in our catalogue, in principle one should be able to compute the spectral variability parameter β using Eq. (2); that is, we are trying to find a correlation between $\Delta\Gamma$ and $\Delta\log F$.

The flux chosen for the search of such correlation is not any of those listed in the catalogue. Many authors (e.g. Paolillo et al. 2004) use typical soft (0.5–2 keV) and hard (2–10 keV) bands, therefore we have computed the integrated flux in the band 0.5–2 keV, which is the sum of the bands 2 and 3; we refer to

http://xraygroup.astro.noa.gr/Webpage-prodec/ thecatalog.html, recently updated to include data from 3XMM-DR5

this as F_S . Following the instructions of Watson et al. (2009), the soft flux F_S is given by:

$$F_S = W_{pn,S} F_{pn,S} + W_{m1,S} F_{m1,S} + W_{m2,S} F_{m2,S},$$
(8)

where $F_{pn,S}$, $F_{m1,S}$ and $F_{m2,S}$ indicate the fluxes received in the soft band S by the three X-ray cameras of the detector. The weights are in turn given by:

$$W_{pn,S} = \frac{\sigma^{-2}(F_{pn,S})}{\sigma^{-2}(F_{pn,S}) + \sigma^{-2}(F_{m1,S}) + \sigma^{-2}(F_{m2,S})},$$
(9)

where $\sigma(F_i)$ indicates the error associated with a given flux F_i .

For a given individual source, the opposite of the slope obtained by a linear fit between Γ and $\log F$ is a good approximation of the spectral variability parameter β . This is not true for an ensemble analysis, in which the average photon indices and fluxes might be very different from source to source. To increase coherence between different sets of observations from different sources, instead of simply computing the linear fit between Γ and $\log F_S$, we have computed such linear fits between the deviations of Γ from the source mean value $\overline{\Gamma}$, as well as the deviations of the logarithm of the flux from its mean value $\overline{\log F}_S$. The slope of this linear fit gives

$$\beta = -0.69 \pm 0.03. \tag{10}$$

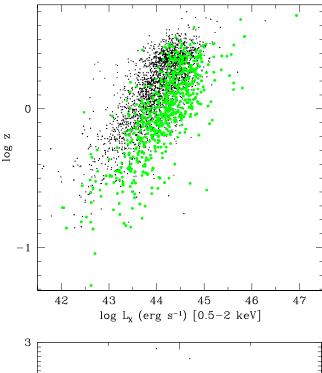
The correlation coefficient is r=0.293, with negligible probability of finding this correlation by chance, $p<10^{-10}$. Given the high heterogeneity of the sample, such moderate correlation and slope in the whole sample are not surprising. The small value of the probability indicates that this correlation, although small, is significant.

In order to give better significance to the data, we also defined a better subsample, in which the worst observations are eliminated, meaning we removed the observations with $\sigma(\Gamma) > 0.5$. Since we are performing spectral fits using only four points, we are likely to obtain significant errors on Γ . This subsample discards 1245 observations, leaving us with 6592 observations of 2309 unique sources. Since we found a value of $\beta = -0.68 \pm 0.02$ for this subsample, no significant difference on the ensemble β was found, and therefore we kept working on the whole sample.

To further validate the consistency of our computing method for the photon indices Γ (see Sect. 3.2), we computed β for the coincident subsample between the XMMFITCAT and our catalogue using both our Γ and the WAPO_GAMMA entry, representing the Γ s obtained by spectral fits assuming an absorbed power-law behaviour of the spectra. We find $\beta=-0.46\pm0.02$ for our Γ values and $\beta=-0.46\pm0.04$ for the Γ values that are present in the XMMFITCAT catalogue. This is further evidence that the calculation of Γ with our method is a good approximation of the actual photon index.

This value is smaller than the one obtained by the whole sample. As is clear from Fig. 2, the XMMFITCAT sample is made up of brighter sources, since they have better spectra. We divided the sample into four bins of the average flux of each source, which are clearly distinguishable in the $z-L_X$ plane, and this means that, on average, the spectrum of brighter sources appears to be less variable than the spectrum of fainter ones. To show this, we computed β for each bin, obtaining a spectral variability parameter that approximately decreases in absolute value with higher average fluxes (see Fig. 3 and Table 2). This is due to the average higher error in the Γ s of fainter flux bins (see Fig. 4).

We stress that these ensemble β s are average values that show how the spectra of these sources vary, on average, with flux. As



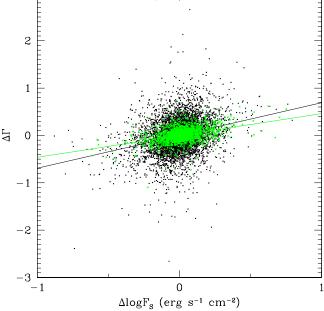


Fig. 2. Top: relation between the X-ray luminosity L_X and the redshift z. Sources in green are the ones that are present in both the XMMFITCAT and MEXSAS catalogue, while the black dots are sources only present in the latter. The green circles are selected by Corral et al. (2015), when the number of counts is at least 50, and these sources are, on average, brighter than the ones not included. Bottom: $\Delta\Gamma - \Delta \log F_S$ correlation, plotted using the photon indices from our analysis, for the same two subsets. The green line shows the linear fit of just the green data, while the black line is the linear fit of the whole sample.

we see in Sect. 3.5, individual sources may even significantly differ from these average values. The negative ensemble value of β indeed confirms a global softer when brighter behaviour, confirming the preliminary result obtained with the hardness ratio method. However, this result does not exclude non-negative values of β , meaning a harder when brighter behaviour for some individual sources, which are indeed present in the sample. We

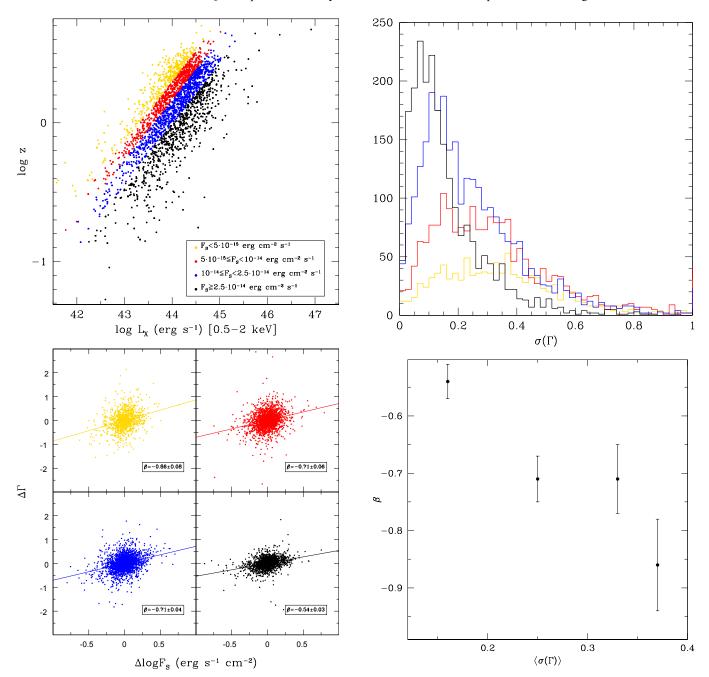


Fig. 3. Top: distribution of the sources in the $z-L_{\rm X}$ plane. Positions of four different flux bins in the $z-L_{\rm X}$ plane are clearly distinguishable as parallel stripes. Different colours represent different average flux bins. Bottom: $\Delta\Gamma - \Delta\log F_S$ plots for the same four flux bins, drawing the linear fit lines for each bin. The slope, which is the opposite of β , decreases for increasing fluxes. The correlation coefficients are approximately 0.3 and are reported in Table 2. All probabilities of obtaining such correlations by chance are $p(>r) < 10^{-10}$.

did not, however, find any such sources among the most significantly correlated ones (see Sect.3.5). In our previous paper (Vagnetti et al. 2016), we estimated the ensemble β of the whole MEXSAS catalogue by means of the structure function, and found the value $\beta = -0.35 \pm 0.02$. Even though the computation of β with the structure function is an indirect method, which was performed using just part of the SF (100–1000 days) and only the three softest *XMM-Newton* energy bands, that estimate agrees with the general trend found in the present analysis.

Fig. 4. $Top: \sigma(\Gamma)$ distribution for the same flux bins defined in Fig. 3. Bins of higher fluxes have smaller average errors and viceversa. *Bottom*: at smaller average Γ errors, we have smaller absolute values of β for the corresponding bins.

3.4. Dependence on physical parameters

As described in Sect. 3.3, our sample is highly heterogeneous. In order to obtain a more detailed analysis, we divided our entire sample into bins of given variables such as redshift, X-ray luminosity, black hole mass and Eddington ratio.

This section demonstrates our method of finding for which values of any one of these given variables the spectrum deviates more or less from the ensemble value, in order to find any possible dependence of the spectral variability parameter on any of these quantities.

Table 2. Four $\langle F_S \rangle$ flux bins with $\beta \pm \sigma(\beta)$, number of sources N_{sour} , number of observations N_{obs} and correlation coefficient r.

$\langle F_S \rangle_{\min}$	$\langle F_S \rangle_{\rm max}$	$N_{\rm sour}$	$N_{\rm obs}$	$\beta \pm \sigma(\beta)$	$\langle \sigma(\Gamma) \rangle$	r
4.26×10^{-16}	5×10^{-15}	401	1039	-0.86 ± 0.08	0.37	0.295
5×10^{-15}	10^{-14}	729	1923	-0.71 ± 0.06	0.33	0.261
10^{-14}	2.5×10^{-14}	963	2689	-0.71 ± 0.04	0.25	0.317
2.5×10^{-14}	4.30×10^{-12}	715	2186	-0.54 ± 0.03	0.16	0.329

Notes. A column showing the average Γ error for each bin is also shown, from which it can be seen that fainter sources also have average greater errors on the spectral slope Γ . The bins at higher fluxes have a smaller β (in absolute value), which means they are less variable in spectrum, but higher correlations are found for these sources. All probabilities of finding the correlations by chance are found to be smaller than 10^{-10} .

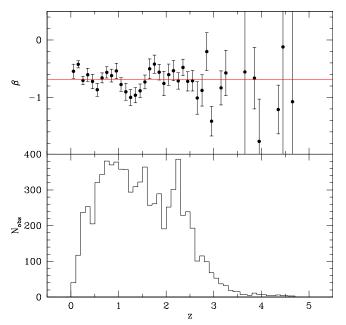


Fig. 5. *Top*: dependence of β on the redshift. The red line represents the ensemble value $\beta = -0.69$. The value of β doesn't show a clear trend with redshift, although some deviations are present for some bins. *Bottom*: redshift histogram of the sample. Up to $z \approx 2.8$, the number of points in each bin is $\gtrsim 10^2$, while for $z \gtrsim 2.8$ the bins include much fewer points, and therefore the β s have greater errors.

3.4.1. Redshift

Both SDSS quasar catalogues DR7Q and DR12Q include the redshift of each source, which is then available for all the sources in our sample.

We divided the whole sample into redshift bins, each containing all the points within a redshift range of size $\delta z = 0.1$. For each bin, we have computed β using Eq. (2).

As can be seen in the upper plot of Fig. 5, there is no clear dependence of β on the redshift. As shown in the bottom plot of Fig. 5, for $z \ge 2.8$ the number of points drops below $\approx 10^2$. Since the fits performed to compute β are less reliable with decreasing number of points, the errors on β are higher.

Thus, we can safely assume that β does not depend on the redshift, therefore the mechanism behind it is not related to the redshift of the source.

3.4.2. X-ray luminosity

It is well known that there is a strong observational correlation between the redshift z and the X-ray luminosity L_X . The location of the sources in the luminosity-redshift plane is shown in Fig. 2.

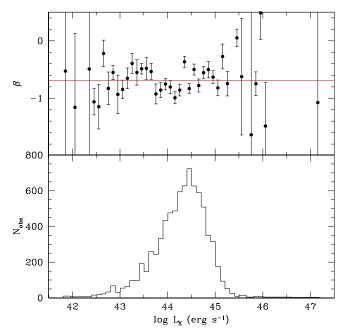


Fig. 6. *Top*: spectral variability parameter β as a function of the X-ray luminosity. The red line represents the ensemble value $\beta = -0.69$. *Bottom*: histogram of X-ray luminosity. The errors on β are very large in those $L_{\rm X}$ intervals where the histogram has $\leq 10^2$ points.

None of the catalogues we used include a list of X-ray luminosities for each epoch, therefore we had to compute it. Using the cosmology introduced in Sect. 1, we compute the luminosity distance of each source as:

$$D_L(z) = \frac{c}{H_0}(1+z) \int_0^z \frac{\mathrm{d}z}{\sqrt{\Omega_{\mathrm{m}}(1+z)^3 + \Omega_{\Lambda}}},$$

and then we compute the X-ray luminosity as:

$$L_{\rm X} = 4\pi F_{\rm X} D_I^2(z) (1+z)^{\Gamma-2}, \tag{11}$$

where we have used the F_S integrated flux between 0.5 keV and 2 keV for the X-ray flux F_X , the redshift of each source and we have adopted a mean value $\Gamma = 1.7$ for every point. The mean value of L_X was then computed for each source. The choice of using a fixed average Γ was taken mostly because the Γ s we are computing are affected by relatively large errors. While the variations $\Delta\Gamma$, and therefore the computation of β , are less affected by these errors, the use of a mean value is preferred for the computing of L_X .

Once we have defined the X-ray luminosity of each source, we grouped the observations in X-ray luminosity bins of size $\delta \log L_{\rm X} = 0.1$ and computed β for each bin, using the same procedure adopted for the redshift. The results are shown in Fig. 6, showing, again, no dependence of β on the X-ray luminosity.

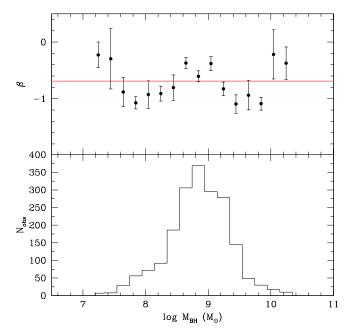


Fig. 7. *Top*: spectral variability parameter β versus black-hole mass $M_{\rm BH}$. Aside from statistical fluctuations, no clear trend is found between the black-hole mass and β . *Bottom*: histogram for black-hole mass.

This should not be surprising, since z and L_X are not independent quantities, as they are strongly related (see Fig. 2). However, it does prove that the spectral variability, and hence the mechanism behind it, does not depend on the X-ray luminosity of the source.

3.4.3. Black-hole mass and Eddington ratio

The MEXSAS catalogue, being derived from XMMSSC-DR5 and from the Sloan quasar catalogues, does not include distinguishing parameters of the source other than the redshift. Moreover, as described in Sect. 3.4.2, we computed the average X-ray luminosity of each source.

To study further parameters, such as the central-black-hole mass and the Eddington ratio of the source, we matched the MEXSAS catalogue with a catalogue containing the properties of 105 783 quasars (Shen et al. 2011). This catalogue³ is limited, however, because it only includes sources that are present on SDSS-DR7Q, and therefore crossing it with MEXSAS means we did not analyse all those sources that are present in the DR12Q but not in the DR7Q. We obtain a subsample made up of 565 unique sources and 1953 observations.

First, we divided this sample into bins of black-hole mass logarithm of $\delta \log M_{\rm BH} = 0.2$ amplitude and computed the ensemble β for each bin. As we can see in Fig. 7, aside from statistical fluctuations from the ensemble value $\beta = -0.69$, a clear trend is not found, and this means that neither the black-hole mass is a driver parameter for the spectral variability.

It should be stressed that the black-hole masses listed in Shen et al. (2011) are single epoch virial estimates computed from the H β , MgII and CIV lines, depending on the redshift of the source. These estimates, especially the masses obtained from the CIV line, are highly uncertain.

We also divided the sample into bins of Eddington ratio $\lambda_{\rm E} = L_{\rm bol}/L_{\rm Edd}$ of $\delta \log \lambda_{\rm E} = 0.5$ amplitude. As shown in Fig. 8,

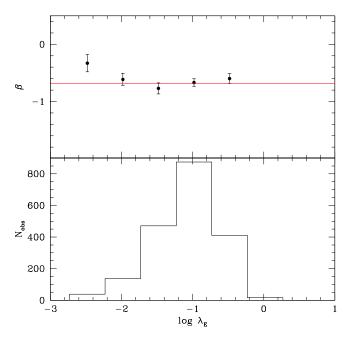


Fig. 8. Top: spectral variability parameter β versus Eddington ratio λ_E . There is no clear trend for $\lambda_E \gtrsim -2$. Even though the number of points for the first bin is scarce, the absolute value of β for very low Eddington ratio $\lambda_E \lesssim -2$ is smaller, suggesting that for even lower Eddington ratios it might become positive, transitioning to harder when brighter trend as found by Connolly et al. (2016). Bottom: histogram of Eddington ratio is also shown.

 β does not depend on λ_E either for $-2 \lesssim \lambda_E \lesssim 0$, while it shows a weak trend toward smaller absolute values for $\lambda_E \lesssim -2$, suggesting a possible agreement with Connolly et al. (2016). In fact, according to their work, there might be a transition to harder when brighter trend ($\beta > 0$) for very low Eddington ratio sources. The trend in our data is not strong enough, given the small number of observations contributing to small λ_E .

3.4.4. BALQSOs and radio-loud sources

As briefly reported in Sect. 2, broad absorpion line quasars (BALQSO) and radio-loud sources are also present in our sample. Both classes might differ in spectral variability properties from the overall trend, due, respectively, to X-ray absorption typically associated with the UV BAL outflows (e.g. Brandt et al. 2000) and to the enhanced X-ray emission associated with their jets (e.g. Worrall et al. 1987).

To quantify the number of BAL sources present in our catalogue, we cross-matched it with the already mentioned catalogue by Shen et al. (2011), which lists a flag to recognise BALQSO based on the catalogue by Gibson et al. (2009). We find 31 observations of 14 sources in the DR7Q subset of our catalogue. As for the DR12Q subset of our catalogue, the quasar catalogue itself lists the BALQSO flag (Pâris et al. 2017) and we obtain 171 observations of 63 sources. Removing 13 observations of 6 sources that are found in both DR7Q and DR12Q, we have a total of 189 observations of 71 unique sources for the merged BALQSO list.

The spectral variability parameter of this subset was then computed and we find $\beta = -0.64 \pm 0.16$, with correlation coefficient r = 0.285 and a probability of finding such correlation by chance given by $p(>r) \simeq 9 \times 10^{-5}$. This estimate is not dissimilar from the ensemble average value of the spectral variability

 $^{^3}$ http://quasar.astro.illinois.edu/BH_mass/dr7.htm

Table 3. Single-source results are shown in this table.

IAUNAME	Serial number	$\beta \pm \sigma(\beta)$	$\langle F_S \rangle$	Nobs	r	p	z	$\log L_{\mathrm{X}}$
3XMM J095847.8+690532	97	-1.57 ± 0.28	7.32×10^{-14}	13	0.861	7.9×10^{-5}	1.288	45.02
3XMM J100043.1+020637	117	-1.17 ± 0.15	3.80×10^{-14}	11	0.932	1.5×10^{-5}	0.360	43.46
3XMM J121753.0+294304	228	-3.54 ± 0.54	3.34×10^{-14}	8	0.938	2.9×10^{-4}	1.647	44.88
3XMM J122931.2+015247	256	-1.43 ± 0.39	1.70×10^{-14}	24	0.613	7.1×10^{-4}	0.770	43.87
3XMM J123759.5+621102	269	-0.81 ± 0.10	5.63×10^{-14}	7	0.964	2.4×10^{-4}	0.910	44.57
3XMM J013943.1+061254	532	-0.80 ± 0.06	5.82×10^{-14}	5	0.992	3.9×10^{-4}	0.678	44.33
3XMM J115535.8+232723	2066	-0.80 ± 0.06	9.24×10^{-14}	5	0.991	4.6×10^{-4}	0.136	42.95
3XMM J121834.6+293453	2125	-1.38 ± 0.11	1.92×10^{-14}	8	0.980	9.3×10^{-6}	1.843	44.81

Notes. To identify each source, we have reported both the 3XMM-DR5 catalogue entry IAUNAME in the first column and the serial number in the MEXSAS catalogue that labels each source, which is shown in the second column. For each source, the spectral variability parameter β and its error, the number of epochs N_{obs} , the correlation coefficient r, the probability of finding such correlation by chance p, the redshift z and the X-ray luminosity $\log L_X$ are shown. Also shown is the average value of the flux $\langle F_S \rangle$, from which we can see that all of the most correlated sources come from the two brightest flux bins (see Table 2).

parameter and, therefore, the BALQSOs do not affect the general ensemble analysis.

We then cross-matched the Shen et al. (2011) and the MEXSAS again in order to find radio-loud sources. We identified 169 observations of 56 unique sources with available radio data, with 137 observations of 47 unique sources that have radio loudness parameter R > 10, defined as

$$R = \frac{F_{6 \text{ cm}}}{F_{2500 \text{ Å}}},\tag{12}$$

after Jiang et al. (2007), that is, they are radio-loud sources. This catalogue only lists data for the sources from the SDSS-DR7Q.

Therefore, we identified the FIRST flux (Becker et al. 1995) at $\lambda=20$ cm in the SDSS-DR12Q catalogue and, in order to have comparable results with Shen et al. (2011), we computed the radio flux at rest-frame wavelength $\lambda=6$ cm using $F_{\nu} \propto \nu^{\alpha}$. Following Gibson et al. (2008) and Vagnetti et al. (2010), we assume a standard $\alpha_{\rm radio}=-0.8$ value for all the sources. We also identified the ugriz fluxes in the SDSS (Fukugita et al. 1996) and computed the flux at 2500 Å by interpolating between the nearest SDSS filters in the rest-frame, or extrapolating in the cases when all the filters lie on the same side with respect to 2500 Å, adopting in such cases $\alpha_{\rm UV}=-0.46$, following Vanden Berk et al. (2001).

Then, we computed the radio-loudness parameter also following Jiang et al. (2007). We found radio data for 244 observations of 80 sources in the SDSS-DR12Q catalogue, which makes a total of 126 sources at 384 epochs if we include the Shen et al. (2011) data for the DR7 sources and exclude nine sources that are available in both catalogues, for which we chose to adopt the radio-loudness as computed by us. In this sample, R ranges from 0.3 up to 5×10^4 . Most of these sources (114) are radio-loud (R > 10) and we find a spectral variability parameter $\beta = -0.38 \pm 0.10$ for this sample. Compared to the result obtained for all the sources (see Eq. (10)), this means that, on average, the spectra of radio-loud sources are less variable than the rest. A similar behaviour is also seen in the optical/UV bands (Vagnetti et al. 2003).

3.5. Individual sources

As shown in Sect. 3.4, the ensemble spectral variability parameter β does not appear to change with the distinctive features of the quasar, such as black-hole mass, redshift, Eddington ratio

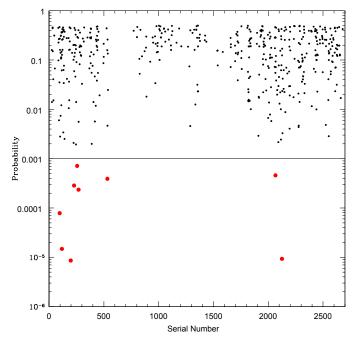


Fig. 9. Probability of finding Γ -log F_S correlation by chance. The most significantly correlated sources are chosen by having probability lower than 10^{-3} (horizontal line). The nine selected sources are shown as red circles. One of these sources (198, also known as PG 1114+445) was discarded due to the well-known presence of a warm absorber, reducing the number of analysed sources to eight.

or X-ray luminosity. However, as is clear from Figs. 5–8, some kind of deviation from the ensemble value is obviously allowed. For some bins of X-ray luminosity, we even find $\beta > 0$, hinting at a harder when brighter behaviour for the sources present in those bins, opposite to the general trend of the whole sample (see Fig. 6).

This means that single sources as well could deviate even significantly from the general trend of the ensemble. To study this, we computed β for each source with $N_{\rm obs} > 3$. We specify that most of these β values are not significant, because they are either not correlated or the fits are performed on too few points to be acceptable.

Many of the sources listed have a very low individual correlation between their own fluxes F_S and spectral indices Γ , which means that they are mildly variable in spectrum or have

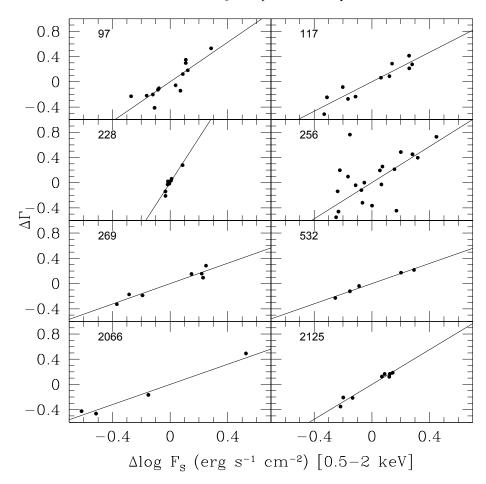


Fig. 10. $\Delta\Gamma - \Delta\log F_S$ plots for eight sources with high correlation and low $(p \le 10^{-3})$ probability of finding such correlation by chance. The slopes are different from source to source, ranging from $\beta = -3.54$ to $\beta = -0.80$, suggesting that even though the mechanism behind spectral variability is likely the same, there is a certain range among the sources. The MEXSAS serial numbers are reported in each panel, see also Table 3.

too few observations to tell. The probability of finding a significant $\Gamma - \log F$ correlation by chance is plotted in Fig. 9 for each source. To estimate this correlation probability, given the small number of points involved, we perform a Student's t-test (Bevington 1969), which computes the integral probability of the null hypothesis p(>r, N) for a given correlation coefficient r with N pairs of data points. We selected those sources for which the probability of finding a casual correlation is very small $(p \le 10^{-3})$, in order to find the most confident results. We have therefore selected nine sources, eight of which are listed in Table 3. Indeed, after visual inspection of the available X-ray spectra, we excluded source number 198, that clearly shows the presence of a warm absorber. This source can be identified with PG 1114+445, a well-known source whose warm absorber has already been studied in the past (e.g. Ashton et al. 2004). In the ensemble study, we estimated that the presence of additional components, such as a warm absorber, only affect 10% of the sources, and therefore can be neglected. However, the presence of these features in a single source study requires a more detailed analysis, which is beyond the scope of this paper. Therefore, we only analysed the eight individual AGNs listed in Table 3. We note that these sources are distributed in a wide range of X-ray luminosities and redshift.

All of them show a softer when brighter behaviour, although to a varying extent, since β ranges from $\beta = -0.80 \pm 0.06$ for sources numbered 532 and 2066, to the extreme value of $\beta = -3.54 \pm 0.54$ for the source numbered 228 in the MEXSAS catalogue (see Fig. 10). All of the sources belong to the two highest-flux bins, confirming that the brightest sources are the most correlated ones.

4. Summary and discussion

We have found an ensemble softer when brighter behaviour for a sample of 2700 quasars by means of the spectral variability parameter β , first introduced by Trevese & Vagnetti (2002). A negative value ($\beta = -0.69 \pm 0.03$) was found, which means that a positive slope between $\Delta\Gamma$ and $\Delta\log F$ is present. This positive slope means that, on average, a softer when brighter trend is present for our sample. As can be seen in Figs. 3 and 4, the errors in $\Delta\Gamma$, which are higher for the sources with lower average flux, give rise to a wider dispersion of the points in the $\Delta\Gamma$ – $\Delta\log F_S$ plane, causing a steeper slope of the linear fit, meaning a larger absolute value of β . Even considering only the sources with the best spectrum (the ones in common with the XMMFITCAT catalogue released by Corral et al. 2015), we still obtain a negative β , slightly smaller in absolute value, $\beta = -0.43 \pm 0.04$. Dividing our sample into average flux bins, the brightest sources give $\beta = -0.54 \pm 0.03$, still confirming a softer when brighter trend.

This result should be viewed in the context of the existing models for the X-ray emission and spectral variability engines.

The current paradigm for X-ray emission involves the presence of two mutually interactive components: a cold, optically thick phase, that is, the accretion disk, that provides input soft UV photons, and an optically thin phase, that is, a hot electron plasma, commonly known as the corona (Liang 1979; Haardt & Maraschi 1991). The soft photons that are emitted by the accretion disk are up-Comptonised by the corona in the X-ray band with a power-law spectrum. Nandra & Papadakis (2001) found that the X-ray changes of the spectral index respond to the UV emission variations, when studying the spectral

variability of the nearby source NGC 7469. Part of the X-ray radiation emitted by the corona gets reflected by the disk and torus (Magdziarz & Zdziarski 1995; McHardy et al. 1999; Shih et al. 2002).

The observed softer when brighter trend might be due to the fact that the soft part of the spectrum is variable in amplitude but not in photon index, while the reflection component is approximately constant. In that case, the variations of the observed photon index Γ_{obs} are due to the combination of these two spectral components, but not to changes of the intrinsic photon index Γ_{int} (see e.g. Taylor et al. 2003; Ponti et al. 2006; Miniutti et al. 2007).

Another option is that the soft component itself is already variable in spectrum, and that the variations of the observed spectral index Γ_{obs} correspond to changes of the intrinsic spectral index Γ_{int} , which increases with decreasing coronal temperature (Haardt et al. 1997).

We also tried to identify different trends for different source features, finding no clear trend between β and the redshift z, the X-ray luminosity $L_{\rm X}$, the black-hole mass $M_{\rm BH}$ or the Eddington ratio $\lambda_{\rm E}$. This suggests that the main mechanism that generates the observed spectral variability trend is likely the same for all sources and probably not due to the physical characteristics of the source. Even though the spectral variability parameter and these quantities are not correlated, there is some deviation from the average value of β on different bins of z, $L_{\rm X}$, $M_{\rm BH}$ and $\lambda_{\rm E}$. We have also computed β for a subsample of 114 radio-loud sources, finding a smaller spectral variability with respect to the general trend. A subset of 71 BALQSOs was also considered, but the spectral variability parameter does not sensibly differ from the ensemble value.

Individual β s were also computed for a restricted set of eight sources, which display the most significant $\Gamma - \log F$ correlations. For these sources, β ranges from $\beta = -3.54 \pm 0.54$ to $\beta = -0.80 \pm 0.06$. It is not clear why the spectral variability parameter, although still maintaining the softer when brighter trend in all cases, is quantitatively different from source to source. One possibility is that they have different orientations and that the spectral variability could be dependent on the angle of view i. Another possibility is a dependence on the black-hole spin, about which there is no available information for these sources. The presence of hidden spectral features, such as soft excess and/or warm absorbers, could also alter the observed value of β . Alternatively, the process could be entirely stochastic, and therefore not correlated to any of the quantities mentioned above.

Future observations of these objects might hopefully shed some light on the topic. For instance, in the near future the mission eROSITA (Merloni et al. 2012) is scheduled to launch. This mission is primarily dedicated to investigating the mass function of galaxy clusters, and during the first four years will perform all-sky searches in order to find 10^4-10^5 such objects. It is estimated (Predehl et al. 2010) that it will be able to detect $\sim 10^6-10^7$ AGNs with redshifts up to $z\sim 7-8$, with an average of eight flux measurements for each source over the four years (Merloni et al. 2012). One such vast catalogue would significantly improve the statistical significance of such variability studies.

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