

# Nonstationarity of AGN variability: The only way to go is down!

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## ABSTRACT

To gain insights into long-term Active Galactic Nuclei (AGN) variability, we analyze an AGN sample from the Sloan Digital Sky Survey (SDSS) and compare their photometry with observations from the Hyper Suprime-Cam survey (HSC) observed  $\langle 14.85 \rangle$  years after SDSS. On average, the AGN are fainter in HSC than SDSS. We demonstrate that the difference is not due to subtle differences in the SDSS versus HSC filters or photometry. The decrease in mean brightness is redshift dependent, consistent with expectations for a change that is a function of the rest-frame time separation between observations. At a given redshift, the mean decrease in brightness is stronger for more luminous AGN and for objects with longer time separation between measurements. The mean decrease in brightness with time violates the stationarity assumption often invoked in AGN variability studies. We demonstrate that the dependence on redshift and luminosity of measured mean brightness decrease is consistent with simple models of Eddington ratio variability in AGN on long (Myr, Gyr) timescales. We show how our results can be used to constrain the variability and demographic properties of AGN populations.

*Keywords:* accretion, accretion disks - black hole physics - methods: data analysis - quasars: general

## 1. INTRODUCTION

Changing flux levels with time are nearly ubiquitous among Active Galactic Nuclei (AGN). Studies of these luminosity fluctuations, i.e., AGN variability, have enabled measurements of central supermassive black hole masses (e.g., [Bentz 2015](#)), added insights on the structure of AGN accretion disks (e.g., [Fausnaugh et al. 2016](#)) and provide powerful AGN selection techniques (e.g., [Schmidt et al. 2010](#)). AGN variability has been directly observed in large samples on timescale ranging from minutes to days, years, and decades (e.g., [MacLeod et al. 2010, 2012](#); [Morganson et al. 2014](#); [Cartier et al. 2015](#); [Caplar et al. 2017](#); [Smith et al. 2018](#)). Through indirect methods and simulations, AGN variability has also been studied on Myr and Gyr scales (e.g., [Novak et al. 2011](#); [Bland-Hawthorn et al. 2013](#); [Sartori et al. 2018](#)).

Most of the direct observational studies mentioned above quantify AGN variability as a weakly stationary process, i.e., with the assumption that the mean luminosity of statistically large ensembles of AGN does not change with time. Empirically, stochastic variability measured in these short-term studies dominated any possible subtle changes in the mean brightness occurring during the duration of the studies. From theoretical grounds, the stochastic variability is thought to be reflective of the details of the physics of AGN accretion disks and other nearby structures, while the mean change of luminosity would be connected to long timescale accretion processes thought to have minimal impact on typical survey timescales (but see [Lawrence 2018](#)).

The assumption of stationarity on short timescales differs from the long term studies of AGN activity, which indicate large changes in AGN activity on Myr and Gyr scales. The firmest observational proof comes from the studies of individual extended AGN photoionised clouds, so-called “Voorwerp” objects (e.g., [Sartori et al. 2016](#); [Johnson et al. 2018](#)), and the He II transverse proxim-

ity effect (e.g., Schmidt et al. 2018), which clearly show that some AGN exhibit order-of-magnitude changes in their luminosity on  $10^4$ - $10^5$  yr time-scales.

There has been comparatively little observational research on the deviations from the assumed stationary and symmetric behaviour of AGN variability. Numerous early studies conducted observationally difficult searches for potential differences in the variability properties of AGN which were becoming brighter or dimmer, but found little or no evidence for statistical differences in the properties of AGN light curves that were fading or getting brighter (de Vries et al. 2003, 2005; Bauer et al. 2009; Voevodkin 2011). MacLeod et al. (2012) combined the earliest statistically significant sample of AGN measurements from the Palomar Observatory Sky Surveys (POSS) with the SDSS data. They noted that objects from POSS are dimmer when observed in SDSS. They concluded that this may be explained by a “Malmquist-like” bias, i.e., the fact that luminosity selected sample of variable objects will necessarily be dimmer in the later survey, even if there is no change in the mean brightness of the underlying sample. A similar conclusion was reached by Rumbaugh et al. (2018) who studied examples of extreme variability by comparing SDSS and Dark Energy Survey measurements. Morganson et al. (2014) also found the decrease of the mean brightness on decade timescales, for the sample of AGN from the SDSS observed in Pan-STARRS1, but attributed this effect to the filter differences.

Here, we use the AGN sample from SDSS and measure their mean brightness in SDSS and HSC. The depth, size, time-separation from SDSS, and quality of the HSC survey makes it especially suitable for this kind of study. In this work, we aim to show that AGN exhibit changes in their mean brightness in a redshift and luminosity dependent manner on the timescales accessible with past (SDSS) and current (HSC) surveys.

The code, data, and additional examples are available at [github.com/nevencaplar/Nonstationarity](https://github.com/nevencaplar/Nonstationarity).

## 2. OBSERVATIONS

### 2.1. Data

To study AGN variability on decade timescales, we identified AGN from the SDSS (York et al. 2000) DR7 Quasar catalog (Schneider et al. 2010) that were also observed later by the Hyper Suprime-Cam (HSC; Miyazaki et al. 2018) Subaru Strategic imaging survey. The SDSS survey used a dedicated 2.5-m (Gunn et al. 2006) telescope at Apache Point Observatory to obtain images in 5 optical bands (ugriz) over a large patch ( $\sim 10000$  deg<sup>2</sup>) of Northern sky. For a presentation of the photometric calibration and selection function of objects, we refer the

reader to detailed discussion in Schneider et al. (2010). The HSC survey is a wide-field optical imaging program being conducted with the 8.2 meter Subaru telescope. The second public data release (made available in May 2019; Aihara et al. 2019) covers around 300 deg<sup>2</sup> overlapping with the SDSS footprint. The HSC data in 5 optical bands (grizy) is sensitive down to  $\approx 26$ th magnitude.

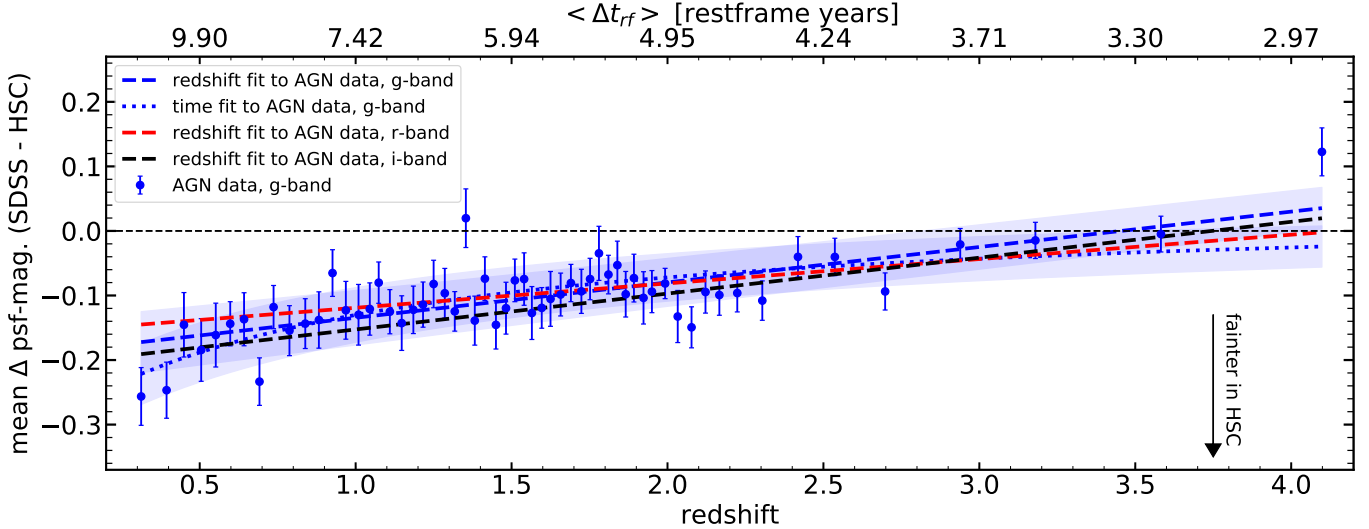
We searched for objects from the SDSS AGN catalog in the HSC data and recorded their  $g$ ,  $r$  and  $i$  psf-magnitudes. We excluded all of the objects with any flags showing problems in the calibration. This conservative cut ensures that our conclusions are not driven by possible problems in the brightness measurements in the HSC pipeline. This procedure yields 5919 matched AGN found in both surveys.

### 2.2. Main result

To measure the mean difference in the brightness between the two surveys, we split the sample in bins of redshift, each consisting of 100 objects. This number enables us to follow the redshift evolution of the trends in some detail, while minimizing the statistical uncertainty in the mean brightness change. For each redshift bin, we then measured the mean and the median difference between the observed psf-magnitudes in the SDSS and HSC surveys. We also verified that our conclusions are unchanged when using fixed 3 arcsec aperture magnitudes.

The resulting mean change in flux is plotted as a function of redshift in Figure 1. To avoid cluttering the plot, we only show data points at each redshift for measurements in the g-band. We estimated uncertainties on the mean value at each redshift by bootstrapping the underlying 100 AGN in each bin. We choose to present g-band variability given that contribution of the host-galaxy light, however small for these bright AGN, will be smallest in the bluest available band. However, results for all three bands are very similar. We also show linear fits to the data in all three bands, where one can explicitly see the similarity between all of the results.

We have also verified that the redshift evolution effect is present if we use median differences instead of mean differences of magnitudes, but the magnitude of the effect is somewhat decreased. For instance, the best fit for the median difference is  $-0.139 + 0.051z$  while for the mean difference it is  $-0.176 + 0.06z$ . The fact that the effect is still present when using the median shows that it cannot be fully explained by a relatively small number of extremely variable quasars (e.g., MacLeod et al. 2016; Rumbaugh et al. 2018). The difference is consistent with a model in which measured changes of the



**Figure 1.** Mean difference in the measured psf-magnitudes for the sample of AGN from SDSS that have been observed in both SDSS and HSC. Blue points show the data for the g-band, while the blue dashed line and the shaded region show the linear fit as a function of redshift to the data and  $1\text{-}\sigma$  uncertainty band. The red and the black dashed lines show the linear fit as a function of redshift in r- and i- bands, respectively. We do not show the data and uncertainty bands for r- and i-band to improve the clarity of the figure, but these are comparable to the g-band quantities. The dotted line shows a fit to the g-band data as a function of restframe time-separation between measurements, indicated on the upper axis.

mean/median flux are the consequence of the long-term AGN behaviour and primarily depend on the restframe time separation between the two measurements. To illustrate this point we also show a linear fit to the data as a function of restframe time separation between two measurements, i.e., as a function of  $14.85 \text{ years}/(1+z)$ , where 14.85 years is the mean time separation between observations (see Section 2.5). This fit also provides a good explanation for the observed data. We discuss the proposed model further in Section 3.

To ensure that the observed redshift dependence is not a spurious artifact due to differences between the two surveys, we conduct four different checks which we list here:

- consideration of filter differences
- constructing a control sample
- separating the AGN sample according to brightness
- separating the AGN sample according to the time separation between the SDSS and HSC observations.

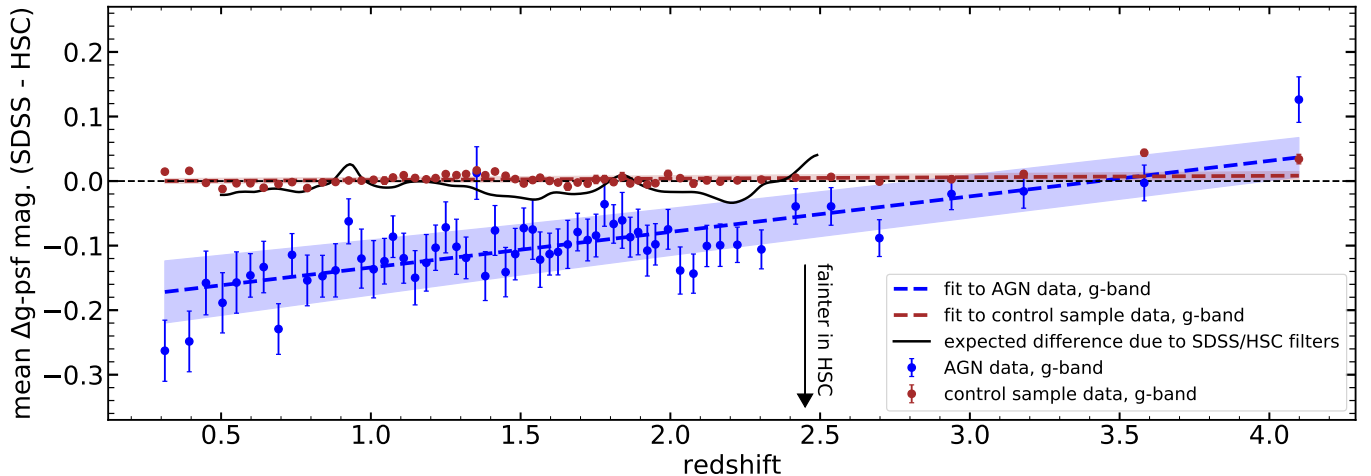
We elaborate each of these procedures in some detail below. For consistency, we always show the mean difference in g-band and conduct linear fits as a function of redshift, but all of our conclusions are applicable to all

three bands and fitting variables (redshift or rest-frame time separation)<sup>1</sup>.

### 2.3. Filter difference and control sample

We performed two experiments to assess the potential impact of differences in the photometry between the SDSS and HSC surveys which could lead to spurious apparent change of measured brightness. First, to assess the potential impact of differences in the filter systems on the measured magnitudes, we predicted  $g_{\text{SDSS}} - g_{\text{HSC}}$  for the mean SDSS quasar spectral energy distribution (Vanden Berk et al. 2001) as a function of redshift using the SDSS (Fukugita et al. 1996) and HSC (Kawanomoto et al. 2018) defined system throughput including the filters, telescopes, cameras, and the survey standard atmospheres. Second, we constructed a control sample consisting of non-variable stars with colors similar to the AGN. The stars were taken from the catalog of non-variable objects from the equatorial Stripe 82 presented in Ivezić et al. (2007) which we additionally cleaned by removing suspected AGN from Flesch (2015). For each AGN we find the star (repetition allowed) which minimizes the euclidean distance between the measured magnitudes in the g-, r- and i- bands from SDSS. After that, we treated the resulting catalog of stars in exactly the

<sup>1</sup> Figures showing results for all of the possible combinations of choices for used observed bands, fitting variables and using mean/median to derive result can be created from the code and the data available in the Github repository



**Figure 2.** Mean difference in the measured psf-magnitudes for the sample of AGN from SDSS, and the control sample of stars that match these AGN in color. The blue points show the data for AGN, while the blue line and the shaded region show the linear fit to the data and 1- $\sigma$  uncertainty. This is equivalent to the data and fit shown in Figure 1. The maroon points, line, and shaded region show equivalent quantities constructed for the sample of non-variable stars in Stripe 82 region. The black line shows the expected redshift dependence due to filter differences between the two surveys.

same way as we have treated the AGN sample. As, by definition, we expect no change in the brightness of these stars when imaged in the two surveys, any systematic differences between the two surveys will be expressed in this comparison. These experiments capture effects both from filter differences and from any differences in the PSF magnitude measurement techniques.

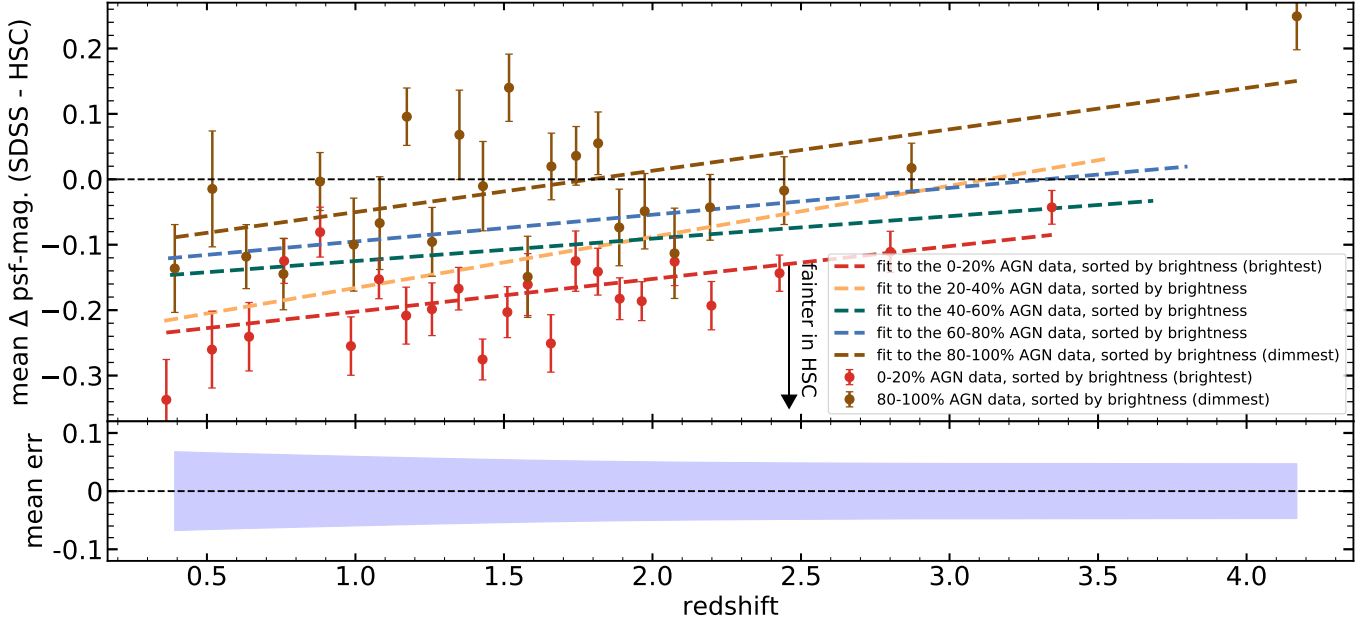
We show the results of this experiment and deduced effects of filter differences in Figure 2. The overall decrease in mean flux is not present in the control sample of non-variable stars. In particular, we emphasize the absence of “redshift” trend in the control sample. This is an expected result, as the different “redshifts” for the control sample correspond to only relatively small changes in the mean color of the objects, which does not affect the calibration of the surveys greatly. We also note that the expected filter differences between the two surveys produce a relatively small and almost redshift-independent effect for the AGN sample. This is due to the small differences between the SDSS (Fukugita et al. 1996) and HSC (Kawanomoto et al. 2018) g-bands,<sup>2</sup> and characteristic power-law SED of an AGN which result in modest  $u-g$  and  $g-r$  colors of  $\approx -0.2$  to  $0.3$  on the AB system (e.g., Richards et al. 2001). Based on these two tests, we conclude that differences in the survey photometry cannot explain the observed difference between two AGN measurements. We proceed with further tests to confirm this conclusion.

<sup>2</sup> with  $\lambda_{\text{eff}} = 4770 \text{ \AA}$  and  $\text{FWHM} = 1379 \text{ \AA}$  for SDSS versus  $\lambda_{\text{eff}} = 4754 \text{ \AA}$  and  $\text{FWHM} = 1395 \text{ \AA}$  for HSC

#### 2.4. Split according to brightness

We then continue to study the redshift effect after splitting our sample in brightness. We do this for two separate reasons. Observationally, we expect that systematic differences between the surveys would be more strongly manifested for objects that have lower brightness, as various errors and uncertainties start to dominate closer to the brightness limit of the SDSS survey. Also, among less luminous AGN, which occur predominantly at lower redshifts, flux from the host galaxy may start to be non-negligible (Shen et al. 2011), which might bias our results. Additionally, given that variability is enhanced at lower luminosities, “Eddington bias”, referring to the fact that intrinsically lower-luminosity AGN might get scattered into the selection of the first shallow survey and then “return” to their mean value when observed later, would produce a measured mean change of brightness which would be more noticeable at lower luminosities.

On the other hand, physically, we would expect that the mean brightness change would be larger for more luminous AGN. Under the assumption that all AGN are the members of the same population, with the same underlying Eddington ratio distribution, AGN are bright enough to be detected in a shallow flux limited-survey only during rare parts of their life-cycle. We would expect that, on average, the population of such AGN would gravitate to their mean, low-flux state as a function of time. In particular, if this assumption is correct, we would expect that brighter AGN are in the more extreme part of their life-cycle, occupying more extreme ends of their long-term Eddington ratio distribution. We



**Figure 3.** Mean difference in the measured psf-magnitudes in g-band for the sample of AGN separated according to their brightness at each redshift. Different colored lines show the results of the linear fit to the subsets of the data. They have been constructed by separating the data at each redshift in quintiles, according to their brightness. Binning is coarser than in Figures 1 and 2 as to preserve statistical power of individual data points. We do not show the data for the 3 inner-quintiles of the data to improve the clarity of the figure. Lower panel shows mean  $1-\sigma$  uncertainty bands around linear fits. Note that scaling on y-axis is different than in the upper panel.

would therefore expect that brighter AGN will decrease their brightness more during any given observed time-frame.

We split the data in each redshift bin into 5 further bins, according to their observed brightness in SDSS. We then proceeded to fit the data in each of these brightness bins with a linear function and show the results of the fitting procedure in Figure 3. As uncertainties on the fits are quite similar for all of the 5 bins, we show the mean error on the fit in the separate panel below the main panel. We see that the effect is indeed stronger for the brighter AGN, as we expected from our theoretical reasoning. This makes us even more confident in the physical nature of this effect. We also wish to point out that the observed brightening for the dimmest objects is mostly driven by the last point at the highest redshift, and it is not obvious that it is also a physical result.

### 2.5. Split according to the time separation

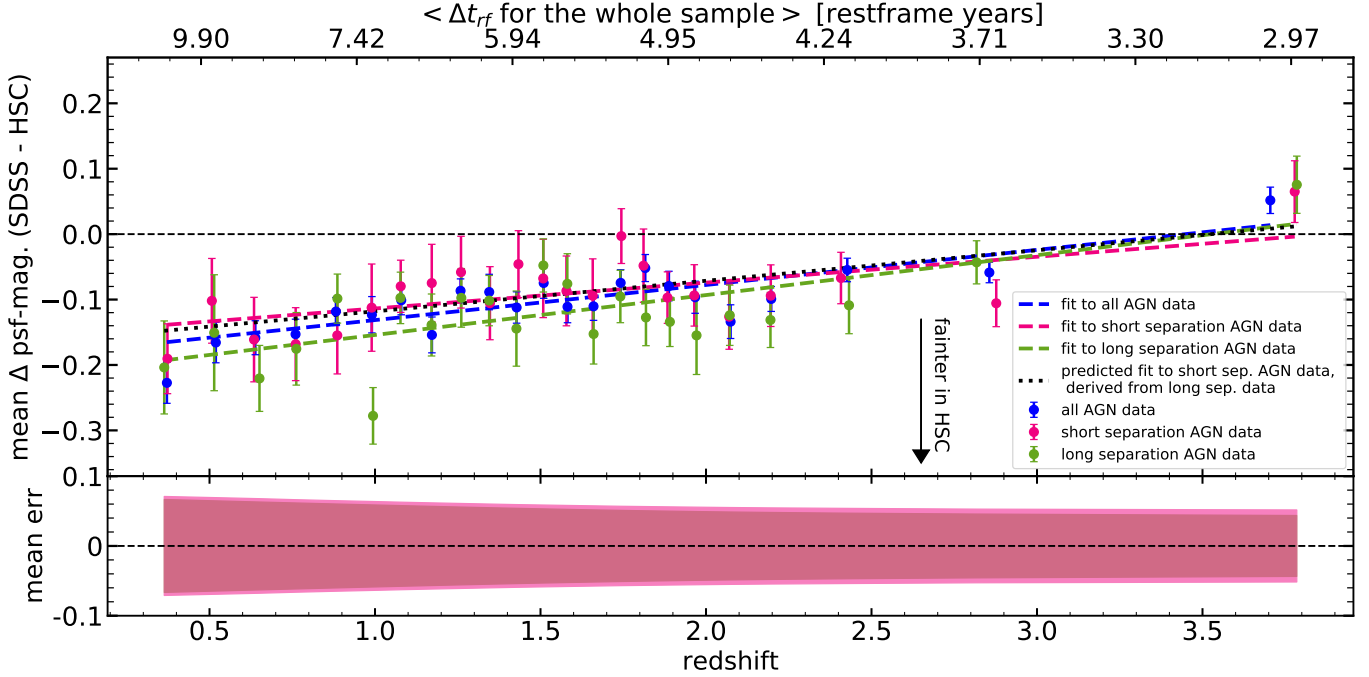
As a final check, we separated our sample in the quintiles according to the time-separation between the observations. As both surveys took data over several years, we split the samples into those taken, by random chance, at shortest and longest time intervals and compare the results. If the change of mean brightness is mostly due to observational effects, we would expect no difference between the short and long separation datasets, while if

the difference is physical we would expect to see some difference between these two sets.

This experiment is somewhat complicated by the fact that we, at this stage, are only working with the stacked HSC data, i.e., the measured brightness of any object is a combination of measurements at different times during the duration of the survey. For HSC data, we take the mean of all of the observation times that go into each stacked observation and use that “mean time” as the time of the observation. The distribution of time differences between the surveys is roughly normal, with the mean at 14.85 years. We then create a sample out of the data in each redshift bin for which the time separation is within shortest time separations quintile (short separation sample) and out of the data which is in the longest time separations quintile (long separation sample). Mean time separation for the short separation sample is 12.94 years, and for the long separation sample is 16.89 years.

We then proceeded as before to study the redshift dependence of each of these samples. We show the data, the results of the linear fit to the full data and long/short separations at Figure 4. We see that, in general, long separation data does indeed tend to show larger changes between the two surveys. Of course, the results are noisier, which is not unsurprising given the underlying stochastic variability. In Figure 4 we also show the ex-





**Figure 4.** Mean difference in the measured psf-magnitudes in g-band for the sample of AGN split according to the time separation between the measurements. The blue points show the data for the whole sample of AGN in the g-band, while the blue line shows the linear fit to the data. This is equivalent to the data and fit shown in Figure 1, although the binning is different to match binning for short and long separation samples (shown in pink and green, respectively). The black dotted line shows the simplest “derivation” of the short separation fit, which has been calculated from the long separation fit by reducing it by the ratio of the mean time-separations for these two samples (12.94, 16.89 years). The lower panel shows 1- $\sigma$  uncertainty bands on these linear fits for the short and long separation data. Note that scaling on y-axis is different than in the upper panel.

pected linear fit for the short separation sample, which was derived from the long separation sample by multiplying the slope with the ratio of mean time separations of each sample, i.e., with the factor 12.94/16.89. This is a simplified assumption, as mean change in brightness is not necessarily linear with time, but we see that the modifications explain well the magnitude of the observed difference.

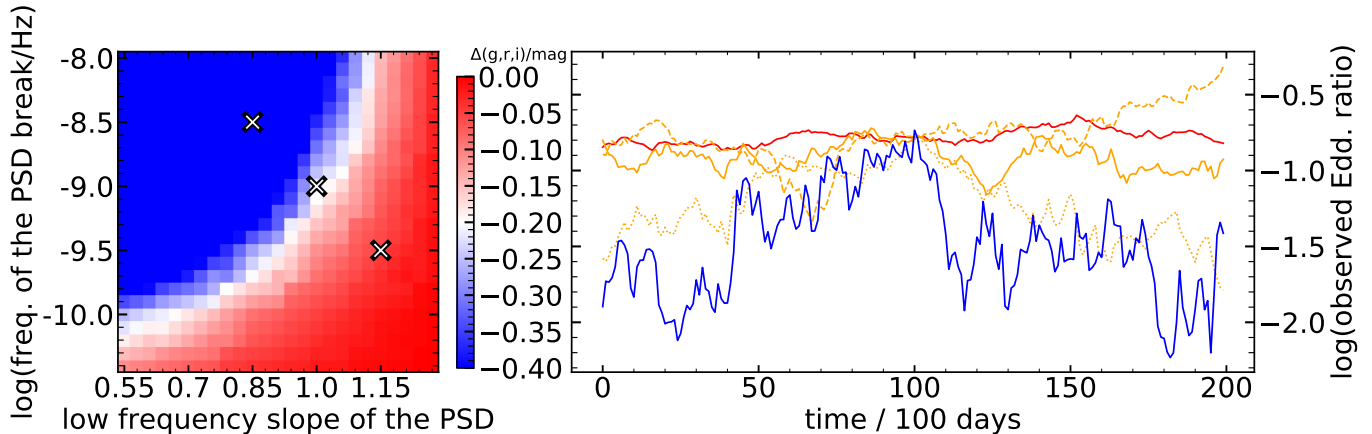
### 3. MODELLING AND DISCUSSION

In this section we discuss how the effect can be used to constrain parameters of AGN accretion given a reasonable set of assumptions. Recently [Sartori et al. \(2019\)](#) developed a code which is capable of simulating Eddington ratio-curves with the duration of Myr to Gyr, and a time resolution of 10-100 days. The inputs to the code are a probability density function (PDF; in this case this is the Eddington ratio function) and the power spectrum density (PSD). We show here an example of how the observed dependence can be used to constrain the PSD parameters. We model the PSD as a broken power, i.e., with

$$PSD(f) = A \times \left[ \left( \frac{f}{f_{br}} \right)^{\alpha_{low}} + \left( \frac{f}{f_{br}} \right)^{\alpha_{high}} \right]^{-1} \quad (1)$$

where  $f_{br}$  is the break frequency, and  $\alpha_{low}$  and  $\alpha_{high}$  are the slopes at lower and higher frequencies, respectively (longer and shorter timescales, respectively). While there is agreement in the community that  $\alpha_{high} \approx 2$  (except perhaps at shortest scales, <10 days, e.g., [Edelson et al. 2014](#)), the deduced values for  $\alpha_{low}$  and  $f_{br}$  vary greatly depending on the survey and method used (e.g., [MacLeod et al. 2010, 2012](#); [Graham et al. 2014](#); [Kozłowski 2017](#)). Physically, the determination of  $f_{br}$  is of great interest as it would provide us with a clue about the physical scale on which the properties of AGN accretion change.

In the left panel of Figure 5 we show the expected mean change of the measured brightness during 14.85 years, the average time difference between two measurements, as a function of  $f_{br}$  and  $\alpha_{low}$ . Changing these parameters effectively changes the “burstiness” of the AGN accretion episodes and therefore influences how quickly the AGN are changing their luminosity in a fixed time period. This plot has been made for the systems selected with an Eddington ratio cut 0.2 dex (0.5 mag) above the break of the Eddington ratio distribution, which broadly mimics the SDSS observational cut. We can see that the observed mean brightness change



**Figure 5.** *Left:* The mean change in measured brightness for AGN sampled at 0.5 mag above the brightness cut of a hypothetical survey, and measured again 10 restframe years later, as a function of  $\alpha_{\text{low}}$  and  $f_{\text{br}}$ . *Right:* Typical “light curves” (observed Eddington ratio) curves, from each of the areas denoted with a small cross on the left-hand side. Colors correspond to the colors on the left-hand side, where we use orange (instead of white) to color the curves from the middle, observationally plausible, region. We show three curves from the middle region to demonstrate the diversity of behaviours.

defines a very specific range of allowed values in this parameter space. The two parameters are somewhat degenerate - observationally, when using limited amount of data points, there is little difference if the process decorrelates quickly at longer timescale (small  $\alpha_{\text{low}}$  and large  $f_{\text{br}}$ ) or slowly at shorter timescales (large  $\alpha_{\text{low}}$  and small  $f_{\text{br}}$  - see also Figure 12. in Caplar & Tacchella 2019). On the right-hand side of 5 we show representative “light curves” from different regions of the parameters space. In actuality we generate curves which satisfy observed Eddington ratio distributions, and we make an assumption that variability in these “light curves” is equivalent to the variability in the observed light curves. In particular, we emphasize the wide variety of the behaviours for the curves which are consistent with the observed changes in the mean brightness. This is reminiscent of the wide diversity of observed variability behaviours for AGN.

Qualitatively, as indicated before, this model also explains why the most luminous objects at lowest redshift are more likely to get dimmer. As they already occupy the uppermost edges of the probability density function (Eddington ratio distribution), they are far more likely to get dimmer and move to more common regions of the

parameter space. In other words, for the brightest AGN, the only way to go is down!

In the future, we aim to improve observational constraints and finely map time dependence by incorporating information from different surveys, such as POSS, Pan-STARRS, GAIA, and Zwicky Transient Factory. We also aim to use this full information and luminosity dependence of the effect to place fine constraints on the evolution of properties (primarily power spectrum density) describing AGN variability.

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