Using principal component analysis to understand the variability of PDS 456

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1 INTRODUCTION

Outflows from active galactic nuclei (AGNs) are a promising candidate for driving AGN feedback (see review by Fabian 2012). Because of the wider opening angle relative to jets, winds from the AGN accretion disc can potentially couple more effectively to galactic gas, regulating the expansion of the host galaxy. The most extreme such outflows are known as ultra-fast outflows (UFOs), and have outflow velocities of 0.03–0.4c. These outflows are most frequently detected using high energy Fe XXV/XXVI absorption lines, found in the 7–10 keV energy range (e.g. Tombesi et al. 2010; Gofford et al. 2013). However, this energy band is towards the edge of the detector range of the current generation of CCD detectors, where the sensitivity declines steeply. This means that just detecting these features is very difficult, so there are very few studies on their variability or timing properties.

Principal component analysis (PCA) is a powerful tool for understanding complex data sets, such as the results of large observing campaigns on AGNs, which show complex and rapidly variable spectra produced by the interactions of multiple spectral components. We have used this technique extensively to explore the variability of AGN over recent years (e.g. Parker et al. 2015). Most recently, in Parker et al. (2017a), we demonstrated that the absorption features from the rapidly variable ultra-fast outflow in the extreme narrow-line Seyfert 1 (NLS1) IRAS 13224-3809 produce peaks in simple variability spectra calculated using data from ~2 Ms of observations. Variability peaks due to the Lyα lines of Fe XXV/XXVI, Ca XX, Ar XVIII, S XV, S XIV, Mg XII and Ne X are clearly present in the combined data set, with a velocity equal to that found using EPIC-pn and RGS spectroscopy (Parker et al. 2017b; Pinto et al. 2017). These variability peaks indicate that the UFO is responding to the source flux, with the equivalent width of the lines dropping as the flux rises. This is most obviously explained by changes in the ionization of outflowing gas in a disc wind, but could potentially also be explained by geometry changes in a scenario where the absorption lines are produced in an absorbing layer on the disc (Gallo et al. 2011, 2013). In this case, the absorption comes from the outer Thomson depth of the disc, giving a high column density at large viewing angles, and the absorption is applied primarily
to the reflection spectrum, so tracks the reflection fraction. A third possibility is that at least part of the continuum variability is driven by changes in column density of the UFO gas, such that electron scattering attenuates the continuum flux as the column rises. This would then produce stronger absorption lines when the observed flux is lower. While having the variability driven by purely column density changes is statistically disfavoured (Pinto et al. 2017), it is likely that lower density gas would also be more highly ionized, so linked column density and ionization variability may be able to explain the observed spectral variability.

The prototypical AGN for studying UFOs is the low-redshift (z = 0.184) quasar PDS 456. This source shows strong and highly variable high-energy Fe absorption features (e.g. Reeves et al. 2009, 2014) from a UFO (\nu \sim 0.24 \alpha). Unlike IRAS 13224-3809, PDS 456 does not show a trivial correlation between the equivalent width of the Fe absorption line(s) and the source flux, most likely due to complicating neutral absorption. Matzeu et al. (2016) demonstrated that the spectral variability within the 2013 Suzaku observing campaign contains a mixture of intrinsic source variability and absorption variability, which may obscure any relation between the UFO and the continuum. However, there is a strong correlation between the source flux and the measured velocity of the outflow (Matzeu et al. 2017), which may indicate that the wind is driven by radiation pressure. A potential correlation between the source flux and the UFO strength in PDS 456 is also discussed in Nardini et al. (2015), in the context of ionization changes.

Because of the high mass of PDS 456 (~10^9 M\odot; Nardini et al. 2015) its variability is slow, occurring largely between, rather than during, observations. PCA is therefore an ideal tool for studying the long time-scale variability, due to its indifference to the gaps between observations. In this paper, we use PCA to reveal the long-term spectral variability of PDS 456. Specifically, we investigate the linked variability of the X-ray source and outflow, as found in IRAS 13224-3809 by Parker et al. (2017b) and Pinto et al. (2017).

2 OBSERVATIONS AND DATA REDUCTION

2.1 X-ray observations

We use all the available data on PDS 456 from the XMM–Newton and Suzaku archives. The full lists of observations are given in Tables 1 and 2, respectively. The majority of the XMM–Newton observations are from the 2013/14 observing campaign, presented by Nardini et al. (2015), and the majority of the Suzaku observations are from the 2013 campaign, presented by Matzeu et al. (2016) and compared with the earlier observations in Matzeu et al. (2017).

Table 1. Summary of XMM–Newton observations used in this work. We show the observation IDs, start dates, exposure times, and the number of 10 ks spectra used in the analysis after the data are reduced and filtered for flaring background.

| Obs. ID      | Start date       | Exposure time (ks) | Intervals |
|--------------|------------------|--------------------|-----------|
| 00411601001  | 2001 February 26 | 46.5               | 4         |
| 05015801001  | 2007 September 12| 92.4               | 8         |
| 0501580201   | 2007 September 14| 89.7               | 8         |
| 0721010201   | 2013 August 27   | 111.2              | 10        |
| 0721010301   | 2013 September 6 | 113.5              | 10        |
| 0721010401   | 2013 September 15| 120.5              | 11        |
| 0721010501   | 2013 September 20| 112.1              | 10        |
| 0721010601   | 2014 February 26 | 140.8              | 11        |

Table 2. Summary of Suzaku observations used in this work. We show the observation IDs, start dates, exposure times and the number of 30 ks spectra used in the analysis after the data are reduced and filtered for flaring background.

| Obs. ID      | Start date      | Exposure time (ks) | Intervals |
|--------------|-----------------|--------------------|-----------|
| 7010560100   | 2007 October 16 | 190.6              | 12        |
| 7050410100   | 2011 April 26   | 125.5              | 8         |
| 7070350100   | 2013 February 21| 89.7               | 15        |
| 7070350200   | 2013 March 3    | 111.2              | 14        |
| 7070350300   | 2013 March 8    | 113.5              | 9         |

Note: aNote that the exposure times take into account Earth occultations, and thus are significantly shorter than the 30 ks timestep multiplied by the number of intervals.

The XMM–Newton data are reduced using the XMM–Newton science analysis software (SAS) version 15.0.0. We use only the high signal to noise EPIC-pn data, which we extract using the EEPROC tool. We filter the data for background flares, and extract source and background spectra from 40 arcsec circular regions, avoiding the outer regions of the pn detector, which are affected by high copper background.

We use the Suzaku data from the two front illuminated detectors, XIS0 and XIS3. We re-run the Suzaku pipeline with the latest calibration files to generate clean event lists, following the data reduction guide. We use XSELECT to extract spectral products, with 100 arcsec radius circular source and background regions, avoiding contaminating sources (including calibration sources). We sum the data for XIS0 and XIS3 using ADDASCASPEC.

3 RESULTS

We use the method described fully in Parker et al. (2014a) to decompose the data. In brief, the data are sliced into a set of spectra at fixed intervals (10 ks for XMM–Newton, 30 ks for Suzaku). We convert these spectra to an array of normalized residuals to the mean spectrum, which we decompose using singular value decomposition (SVD). SVD returns a set of orthogonal principal components (PCs), which describe the original data as efficiently as possible, along with their corresponding light curves and contributions to the total variance of the data. We calculate the errors using the Monte Carlo method of Miller, Turner & Reeves (2008). We perturb the input spectra with Poisson noise and rerun the analysis 10 times, then calculate the standard deviation in the results.

We analyse the data from the two instruments separately, as they appear to sample different variability regimes (e.g. Matzeu et al. 2016). Unless otherwise specified, all plots are in the rest frame of the source.

3.1 XMM–Newton

In Fig. 1, we show the log-eigenvalue (LEV) diagram for the XMM–Newton analysis. This shows the fraction of the total variance attributable to each spectral component. Those which are caused by noise should follow a simple geometric decay on this plot. We find that three components are not consistent with such a decay, and focus our analysis on them.

1 http://heasarc.gsfc.nasa.gov/docs/suzaku/analysis/

2 Code available from http://www-xray.ast.cam.ac.uk/mlparker/PCA.tar.gz.
Figure 1. LEV diagram showing the fractional variability of each component. The three filled circles are those we identify as highly significant, and which show detailed structure in their spectra. The remainder are consistent with noise. The black line shows a simple geometric decay fit to the higher order components.

These components are shown in Fig. 2. The first component, PC1, is relatively flat, with a strong peak at \( \sim 9 \) keV, corresponding to the energy of the UFO line, and a suppression of the variability at low energies, corresponding to the soft excess. The low energy suppression is frequently found in AGN (Parker et al. 2015), and indicates that the soft excess is less variable than the primary power-law continuum. This also means that the soft excess must be a real emission component, separate from the main continuum and less variable. While it is possible that the lack of variability could be due to a less variable reflection component (where the variability is due to changes in the coronal size or geometry, see e.g. Miniutti et al. 2003; Miniutti & Fabian 2004), the absence of a corresponding iron line feature argues against this interpretation and suggests that the soft excess is due to a separate source of X-ray emission. The peak in variability at the UFO line indicates that the flux in this band is significantly more variable than the continuum, and that it is correlated with the continuum variability. As in the case of IRAS 13224-3809 (Parker et al. 2017a), this is most likely due to ionization of the outflowing gas by the increased source flux.

PC2 is also fairly flat, with a strong anticorrelation between low and high energies. This is due to spectral pivoting – as the low-energy flux rises, the high-energy flux drops, and vice versa. Physically, this behaviour can be produced by changes in the photon index, or by small changes in the absorbing column of neutral gas along the line of sight to the source. We note that Nardini et al. (2015) found evidence that the intrinsic photon index was variable during the 2013/14 \textit{XMM–Newton}/NuSTAR observing campaign (\( \Gamma = 2.1–2.6 \)).

The third component is more complex. There is a broad peak around 6 keV, and a soft excess feature, which are both anticorrelated with intermediate energies. Qualitatively similar variability has been seen in sources with strong reflection components in the past, but there are several factors that mean we cannot robustly attribute this variability to a specific physical mechanism (see Section 4). Interestingly, this component is only strongly present in the first observation, dropping drastically before the second observation and never recovering to its original strength.

To examine the changes between observations 1 and 2 in more detail, we examine the full spectrum of each observation. We fit the two spectra with a power law and Galactic absorption from...
0.3–10 keV (in the observer’s frame) and plot the data/model ratio in Fig. 3. As expected from the spectral shape of PC3, in the first observation there is significant additional emission below 1 keV and from 3–9 keV. We discuss the nature of this variability more in Section 4.

3.2 Suzaku

In contrast to the XMM–Newton data, the spectral variability observed by Suzaku is dominated by variable cold absorption (Matzeu et al. 2016). In particular, the three observations from a 2013 observing campaign show greatly increased absorption, with the flux below \( \sim 5 \) keV dropping dramatically.

The two significant PCs returned from this are shown in Fig. 4, and are clearly showing a different variability pattern from those seen with XMM–Newton, with a dominant soft component and a less variable hard component, which could potentially be closely related to the first two components from XMM–Newton, but in reverse order (we discuss this further in Section 4). PC1 is very soft, consistent with the components observed in sources dominated by absorption variability (see e.g. NGC 4395 in Parker et al. 2015). This is mostly likely caused by a change in covering fraction or column density of the cold gas causing the absorption observed in the spectrum.

3.3 Simulations

To further investigate the UFO variability seen with XMM–Newton, we perform simulations as in Parker et al. (2015): by simulating a set of fake spectra based on simplified models and analysing them with PCA, we can compare to the results with real data and identify the likely causes of the observed variability.

We construct a simple model for the variable spectrum of PDS 456: a constant blackbody at low energies to mimic the less variable soft excess, a power-law varying in normalization for the continuum and a Gaussian absorption line (BBODY+GABS*POWERLAW in XSPEC).

We consider three cases, where the absorption line is correlated, anticorrelated and uncorrelated with the continuum flux. We vary the power-law normalization randomly between 0.5 and 2, and then fix the Gaussian normalization to be proportional, inversely proportional or vary independently over the same range, respectively (for the power-law parameters used, this gives a comparable equivalent width of the Gaussian line to that observed).

The results of these simple simulations are shown in Fig. 5. The only difference is at high energies, where the absorption line feature strongly depends on the source behaviour. If the equivalent width of the line is inversely proportional to the continuum flux, a strong positive feature is visible at 9 keV, as observed in the data. Alternatively, if the equivalent width is correlated with or independent of the continuum, the line feature is negative or absent, respectively. We also show in the right panels of this figure the corresponding \( F_{\text{var}} \) spectra, which are generally similar but do not qualitatively distinguish between anticorrelated and uncorrelated variability.

We also test the effect of allowing the velocity of the absorption line to respond to the source flux, instead of the equivalent width.
emission ionizing the wind, as discussed in Nardini et al. (2015). Parker et al. 2017b), and is most likely caused by the increased X-ray line and the continuum flux observed in IRAS 13224-3809 (Parker et al. 2015), which we identified as being due to relativistic reflection, with the enhanced soft excess and iron line emission caused by an increase in the reflection fraction. However, in this case there are some key differences which prevent us from confidently reaching the same conclusion. First, and most importantly, this component is only seen in one observation, and is not strongly variable within that observation. In the other sources where this is observed, PC3 is present, and variable, in multiple observations, so we can be confident that it is not due to inter-observation variability. In the other sources where this is observed, PC3 is present, and variable, in multiple observations, so we can be confident that it is not due to inter-observation variability, and that the soft excess and iron line features are correlated. Because we only see this component in one observation in PDS 456, we cannot be confident that the increase in soft flux and iron band flux are not coincidental. Secondly, the PC1 does not show the suppression around the iron line associated with relativistic reflection seen in other sources. This deficit is usually seen at the peak of the iron line, where the reflection component is less variable than the primary continuum (presumably due to light-bending effects), damping out the continuum variability. No such feature is visible in PC1 of PDS 456, although the soft excess damping is clearly evident. We note that the 7 keV broad emission present in the spectrum has previously been interpreted as both relativistic reflection (e.g. Walton, Reis & Fabian 2010) and scattered emission from the wind (e.g. Nardini et al. 2015). We note the recent detection of a reverberation lag in PDS 456 by Chiang et al. (2017), which is strong of the line responding. This could potentially mimic the effect of the equivalent width responding, by blue- or red-shifting the line into a different band as the flux changes. As this is not a type of variability that can be described by additive spectral components, PCA is not very sensitive to it. For this, we vary the energy between 8 and 10 keV, proportionally to the log of the flux. This does produce a peak at the mean energy of the feature in PC1, but also gives a sharp drop above this feature. Additionally, it produces a series of components with peaks at different energies within the defined range, so that the feature can be reproduced. For the more moderate variability and smaller velocity shifts observed in the real data, the main effect of this will be to broaden the observed line width. The typical energy shift (\(\Delta E \sim \pm0.5\) keV) is well within the width of the line feature in the PCA spectrum, which stretches from 8–10 keV. It is likely that some correction components are produced to correct for energy shifts, but they presumably contribute little to the total variance and are lost in the noise components.

4 DISCUSSION

The most striking aspect of our results is the detection of the 9 keV absorption feature in the XMM–Newton data with PCA. The UFO line is significantly more variable than the continuum, and also inversely correlated with the continuum flux. This is very similar to the strong anticorrelation between the equivalent width of the UFO line and the continuum flux observed in IRAS 13224-3809 (Parker et al. 2017b), and is most likely caused by the increased X-ray ionization emitting the wind, as discussed in Nardini et al. (2015). In Fig. 6, we show a comparison of PC1 from PDS 456 and IRAS 13224-3809. The Fe peak is sharper and stronger in PDS 456, but there are no significant lower energy features. The most obvious reason for this is that the ionization of the outflow in PDS 456 is higher, and the range of fluxes is much smaller, so the ionization does not drop low enough to produce the lower energy lines. Similarly, at lower fluxes in IRAS 13224-3809 the ionization is low enough that the Fe xxv line starts to weaken again, which weakens the correlation between its equivalent width and the source flux, making the variability peak weaker in comparison to PDS 456. There are lower energy absorption features from relativistic outflowing gas at a lower ionization detected in the XMM–Newton RGS data (Reeves et al. 2016), but these are presumably due to different gas from that detected here, and are difficult to detect in the EPIC-pn data. A note of caution is that PDS 456 and IRAS 13224-3809 are sampled on very different effective time-scales, due to the 2–3 orders of magnitude higher mass of PDS 456. This should not have a major impact on the UFO features, as the recombination time of the gas is very short in the case of IRAS 13224-3809, due to the high densities involved, and still much shorter than the dominant inter-observation variability in PDS 456.

The spectral shape of the XMM–Newton PC3 is very similar to that found in MCG–6–30–15 (Parker et al. 2014a), IH 0707–495 and others (Parker et al. 2015), which we identified as being due to relativistic reflection, with the enhanced soft excess and iron line emission caused by an increase in the reflection fraction. However, in this case there are some key differences which prevent us from confidently reaching the same conclusion. First, and most importantly, this component is only seen in one observation, and is not strongly variable within that observation. In the other sources where this is observed, PC3 is present, and variable, in multiple observations, so we can be confident that it is not due to inter-observation variability, and that the soft excess and iron line features are correlated. Because we only see this component in one observation in PDS 456, we cannot be confident that the increase in soft flux and iron band flux are not coincidental. Secondly, the PC1 does not show the suppression around the iron line associated with relativistic reflection seen in other sources. This deficit is usually seen at the peak of the iron line, where the reflection component is less variable than the primary continuum (presumably due to light-bending effects), damping out the continuum variability. No such feature is visible in PC1 of PDS 456, although the soft excess damping is clearly evident. We note that the 7 keV broad emission present in the spectrum has previously been interpreted as both relativistic reflection (e.g. Walton, Reis & Fabian 2010) and scattered emission from the wind (e.g. Nardini et al. 2015). We note the recent detection of a reverberation lag in PDS 456 by Chiang et al. (2017), which is strong...
evidence for at least some relativistic reflection in the spectrum. We do not find clear evidence in favour of either interpretation using PCA, and further work is needed to understand how and if the two can be distinguished with this method in PDS 456.

There is some variability in PC2 after the first interval, but it is much weaker and we cannot rule out it being due to an entirely different mechanism. We show in Fig. 7 a comparison of PC3 from the full data set and with observation 1 removed, from which it is obvious how much observation 1 dominates the signal in this component. The noise increases dramatically when it is removed, and much of the structure is lost. This shape could potentially be due to relatively small changes in ionized absorption.

We previously used PCA to investigate the variability of PDS 456 as part of a sample of sources in Parker et al. (2015). When that work was written, only the first three XMM–Newton observations (from 2001 and 2007) were publicly available, and the results obtained differed significantly from those we find here. In particular, the variability of the first three observations is dominated by the large spectral changes between observations 1 and 2. This is largely associated with PC3 in this work, which is almost identical to PC2 from Parker et al. (2015). However, the fractional variability is much higher for this component in that work: 9 per cent, compared to 3 per cent here. PC1 from Parker et al. (2015) is significantly different from that found in this work – instead of a flat power law, suppressed at the soft excess and enhanced at the UFO line, the variability was suppressed at intermediate energies, and stronger around the energies of the iron band and soft excess, with no clear signal at the UFO energy. This is also due to the effect of the huge change between the first two observations.

It is interesting that we do not find a strong UFO signature in either of the components from the Suzaku decomposition. Matzeu et al. (2016) found extensive changes in the equivalent width of the 9 keV line between these observations, so we would expect to see some evidence of it in the variability. One possibility is that the line strength is correlated with the source flux (as in the XMM–Newton data), but we cannot cleanly isolate that as a component due to the dominant low ionization absorption variability. Alternatively, the UFO strength could be correlated with both the low ionization absorption column/covering fraction and the intrinsic flux, in such a way that our analysis does not find it to be correlated with either individually. These two components are (at least superficially) very similar to those found from simulations of partial covering absorption where the covering fraction and intrinsic flux both change (Parker et al. 2015, fig. 12), which would argue for the intrinsic inter-

A weakness of this kind of analysis is that it is not sensitive to shifts in energy, as it assumes simple linear additive components. The velocity of the outflow in PDS 456 has been shown to be strongly correlated with the X-ray luminosity (Matzeu et al. 2017), which could potentially impact our results. It could, for example, introduce a spurious correlation between the equivalent width and the flux if the line is blueshifted to a different energy band at higher fluxes. We explored this possibility in Section 3.3, and found that if this is the dominant form of variability on the time-scale probed then it produces a qualitatively different series of PCs than observed.
Additionally, the equivalent widths in Fig. 8 take into account any changes in centroid energy, giving an independent check that the equivalent width is anticorrelated with flux. Energy shifts are clearly present in the data presented by Matzeu et al., but they do not appear to be strongly affecting our results, or even being picked up by our analysis (except possibly as a broadening of the UFO feature). It is likely that there are higher order PCs which describe this variability, but they are lost in the noise.

5 CONCLUSIONS

In this paper, we present the detection of the PDS 456 UFO using principal component analysis. This is the second such outflow to be picked up with this technique, which represents a promising method for detecting X-ray outflows in large data sets or rapidly variable, high signal data. Our main conclusions are summarized here:

(i) We find a strong peak in the variability spectrum of PDS 456 at the energy of the Fe absorption feature produced by the powerful outflow in this source. This indicates that, as in the case of IRAS 13224-3809 (Parker et al. 2017b), the ionization of the relativistic gas is responding to the source flux.

(ii) Unlike IRAS 13224-3809, we do not find any signatures of lower energy (1–6 keV) absorption lines. This is presumably due to the higher ionization of the gas in PDS 456, which is a quasar rather than an NLS1, and the absence of a sufficiently large drop in flux in PDS 456, which means these ions cannot recombine.

(iii) There is a clear qualitative difference between the XMM–Newton and Suzaku data. The variability observed with XMM–Newton is dominated by the continuum, which the UFO line responds to. The Suzaku variability, on the other hand, is produced by neutral absorption, which does not show a clear relationship to the UFO feature.

(iv) We demonstrate using simulations that PCA can distinguish trivially between UFO features that are correlated, anticorrelated and uncorrelated with the continuum. However, it is much harder to detect UFO features with PCA when they are uncorrelated with the continuum.

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