Dynamical Modeling of the CIV Broad Line Region of the $z=2.805$ Multiply Imaged Quasar SDSS J2222+2745

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Abstract

We present the first ever models of the broad line region (BLR) at the peak of active galactic nucleus (AGN) activity using the multiply imaged $z=2.805$ quasar SDSS J2222+2745. The modeled data consist of monthly spectra covering the broad CIV emission line over a 5.3 yr baseline. The models suggest a thick disk BLR that is inclined by $\sim40^\circ$ to the observer’s line of sight and with an emissivity weighted median radius of $r_{\text{median}} = 33.0^{+17.4}_{-12.1}$ light days. The kinematics are dominated by near-circular Keplerian motion with the remainder inflowing. The rest-frame lag one would measure from the models is $r_{\text{median}} = 36.4^{+1.8}_{-1.8}$ days, which is consistent with measurements based on cross-correlation. We show a possible geometry and transfer function based on the model fits and find that the model-produced velocity-resolved lags are consistent with those from cross-correlation. We measure a black hole mass of $\log_{10}(M_{\text{BH}}/M_\odot) = 8.31^{+0.57}_{-0.65}$, which requires a scale factor of $f_{\text{model,0}}(r_{\text{mean,0}}) = 0.20^{+0.09}_{-0.07}$. This is the most precise $M_{\text{BH}}$ measurement for any AGN at cosmological distances and it demonstrates that the precision required for BH-host coevolution studies is attainable.

Unified Astronomy Thesaurus concepts: Strong gravitational lensing (1643); Active galactic nuclei (16); Supermassive black holes (1663); Quasars (1319); Astronomy data modeling (1859)

1. Introduction

Precise measurements of supermassive black hole (BH) masses across cosmic time are a necessary ingredient for understanding BH formation and growth and the connection between BHs and their host galaxies (Ding et al. 2020; Ferrarese & Merritt 2000; Gebhardt et al. 2000). The most successful technique for measuring $M_{\text{BH}}$ outside the local universe is reverberation mapping (RM; Blandford & McKee 1982; Peterson 1993, 2014; Ferrarese & Ford 2005), which measures the response of the broad emission-line region (BLR) to changes in the continuum. By combining emission-line widths with the time lag between continuum and emission-line fluctuations, one can obtain a virial estimate of the black hole’s mass:

$$M_{\text{BH}} = f \frac{c \tau (\Delta V)^2}{G},$$

(1)

The scale factor, $f$, accounts for the geometry, kinematics, and orientation of the BLR, which are intrinsic to each individual region and, in general, unknown. This is the largest source of uncertainty in RM $M_{\text{BH}}$ measurements, estimated to be $\sim0.4$ dex (Park et al. 2012).

Direct modeling of the BLR (Brewer et al. 2011; Pancoast et al. 2011, 2012) avoids this issue entirely by including $M_{\text{BH}}$ as a free parameter. Additionally, models inform us of the structure and kinematics of the BLR and can provide $f$ values for individual BLRs. This opens the possibility of finding correlations between $f$ and other observables that could be used to improve $M_{\text{BH}}$ measurements for all active galactic nuclei (AGNs), not just those with data suitable for modeling (Williams et al. 2018; L. Villafaña et al. 2021, in preparation).

Until now, the dynamical modeling approach has been applied only to nearby AGNs with $z < 0.1$, and only one of these analyses (NGC 5548; Williams et al. 2020) examined the UV-emitting BLR. This is due to the complexities of high-$z$ reverberation mapping campaigns paired with the high signal-to-noise ratio (S/N) spectra required for modeling. High-$z$ $M_{\text{BH}}$ measurements rely on the UV lines, though, since the optical lines are shifted out of the observable wavelength range. Studies of BH growth require precise $M_{\text{BH}}$ measurements across all stages of the universe, so an understanding of the UV BLR in the early universe and in quasar-like environments is necessary.

The extraordinary data set of the multiply imaged quasar SDSS J2222+2745 (discovered by Dahle et al. 2013) described by Williams et al. (2021, hereafter Paper I) is the first high-$z$ monitoring campaign with data quality good enough for the modeling approach. In this paper, we model the C IV-emitting BLR in SDSS J2222+2745 using the data presented in Paper I. In Section 2, we briefly describe the data set and the BLR modeling approach used in the analysis. In Section 3, we present the model fits to the data and describe the inferred BLR geometry and kinematics. We also compute the scale factor $f$ for the SDSS J2222+2745 C IV BLR and compare it to values for

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we adopt a $\Lambda$CDM cosmology with $H_0 = 70 \text{ km s}^{-1} \text{ Mpc}^{-1}$, $\Omega_M = 0.3$, and $\Omega_{\Lambda} = 0.7$. 

2. Data and Methods

The data used in this analysis are the same data presented in Paper I, and the modeling approach is described in detail by Pancoast et al. (2014). Here, we briefly summarize both components, but direct the reader to the respective papers for detailed explanations.

2.1. Spectroscopic and Photometric Data

Beginning in 2016 June we obtained monthly spectra of the three brightest images of SDSS J2222+2745 with the Multi-Object Spectrograph at Gemini Observatory North (GMOS-N; Hook et al. 2004). The spectra covered $\sim$5000–8200 Å after dithering and flux calibration, covering the C IV and C III broad emission lines. In addition, we obtained $g$-band photometry with roughly twice-per-month cadence, beginning in September 2011 with the Alhambra Faint Object Spectrograph and Camera (ALFOSC) at the 2.56 m Nordic Optical Telescope (NOT).

Since we are modeling only the emission of the C IV BLR, we first need to isolate the C IV broad emission line from the other spectral components. As described in Paper I, we model C IV as a fourth-order Gauss–Hermite polynomial, and we use those fits in this analysis. We combine the time series of spectra for each image by first multiplying the fluxes by the corresponding image magnification and then shifting the times by the measured time delays ($\Delta \tau_{AB} = -42.44$ days and $\Delta \tau_{AC} = 696.65$ days; Dyrland 2019), setting image A as the reference. The 2020 spectra for the leading image C suffer from low S/N as a result of the small image magnification paired with relaxed observing condition constraints in 2020 due to the COVID-19 pandemic. For this reason, we remove these data from the analysis.

To speed up the modeling code, we also combine spectra that fall within 7 days of each other (observed frame), after accounting for gravitational time delays. This corresponds to $\sim$2 days in the rest frame, which is over an order of magnitude smaller than the expected BLR size. Since our smooth BLR model would be unable to resolve variations on this timescale, we lose no constraining power by combining these spectra. After making these changes, our final data set consists of 63 spectra covering a 1944 day (5.3 yr) baseline, after accounting for gravitational time delays.

Finally, we bin the individual spectra by a factor of 8 in wavelength to further decrease computation time, giving 93 wavelength bins from 5503.5 to 6239.5 Å. The smooth BLR model is unable to resolve emission-line structure on the 1 Å pixel$^{-1}$ (50 km s$^{-1}$ pixel$^{-1}$) scale of the raw data, and we are using smooth Gauss–Hermite fits to C IV, so attempting to fit all pixels would introduce an unnecessary computational burden while providing no additional constraining power.

2.2. BLR Model

We model the BLR emission using a collection of point particles surrounding the central black hole and ionizing source. The ionizing light propagates outwards and as it reaches the particles, they instantaneously reprocess the light and emit it toward the observer in the form of emission lines.

The positions and velocities of the particles are determined by a number of free parameters, described in detail by Pancoast et al. (2014). To summarize, the particles are distributed in a thick disk with half-opening angle $\theta_0$, and inclined relative to the observer’s line of sight by $\theta_i$ ($\theta_i = 0^\circ$ is face-on). An additional parameter, $\gamma$, determines if the particles are distributed uniformly throughout the thick disk ($\gamma = 1$) or if they are concentrated at the faces of the disk ($\gamma = 5$). The distances of the particles from the origin are drawn from a gamma distribution with shape parameter $\beta$, and mean $\mu$, that has been shifted from the origin by a minimum radius $r_{\text{min}}$.

A parameter $\kappa$ determines the relative brightness of each particle by controlling if particles re-emit preferentially back toward the origin ($\kappa = -0.5$), isotropically ($\kappa = 0$), or away from the origin ($\kappa = 0.5$). The disk midplane can range from being fully opaque ($\xi = 0$) to fully transparent ($\xi = 1$).

A fraction of the particles, $f_{\text{flow}}$, are assigned to have near-circular orbits with radial and tangential velocities drawn from a distribution centered on the circular velocity. The remaining particles all have either inflowing or outflowing trajectories based on the binary parameter $f_{\text{flow}}$ (<0.5 inflow, >0.5 outflow). The radial and tangential velocities are drawn from a distribution centered on the radial escape velocity that is rotated by an angle $\theta_0$ in the $v_r$-$v_\phi$ plane. This allows for particles with purely radial motion ($\theta_0 \sim 0^\circ$) or on highly elliptical, bound orbits ($\theta_0 \sim 45^\circ$).

When interpreting the model parameters, it is important to keep in mind that we are modeling the BLR emission rather than the underlying gas. We do not include in our model the complex photoionization process, which would require additional assumptions about the BLR environment—such as the gas density, temperature, and metallicity distributions—and the relation between the observed $g$-band continuum and the ionizing spectrum.

2.3. Fitting the Model to Data

Using the observed continuum light curve, we can compute the time series of emission-line spectra that a given model produces by summing the contributions of each BLR particle, taking into account the position-induced time lag and velocity-induced wavelength shift. Our goal is to explore the model parameter space such that the model-produced spectra best fit the observed spectra. We assume that the ionizing continuum follows the observed $g$-band continuum and use Gaussian processes as a way to flexibly interpolate between observed data points.

We use a Gaussian likelihood function to compare the emission-line spectra with the model spectra. In post-processing, we also introduce a statistical temperature, $T$, that softens the likelihood function, effectively increasing the spectra uncertainties. This accounts for underestimates of uncertainties as well as the challenge of fitting a complex BLR with a simple model. To explore the parameter space of both the BLR model and continuum model, we use the diffusive nested sampling code DNEST4 (Brewer & Foreman-Mackey 2018). Diffusive nested sampling is a modification of the nested sampling technique that is particularly efficient at exploring complex, high-dimensional probability spaces. From the code output, we produce a posterior sample from which we can infer the model parameter values.
Figure 1. From top to bottom, panel 1: time-series of observed spectra. Panel 2: one possible model fit to the observed spectra. Panel 3: normalized residual (Data − Model)/Data uncertainty. Panel 4: observed spectrum from one of the epochs with uncertainties multiplied by \( \sqrt{T} \) (black points) and the spectrum produced by the model shown in panel 2 (red). Panel 5: integrated C IV emission-line flux of the data (black points) and model (red). Panel 6: observed \( g \)-band continuum light curve (black) and the model fit to the light curve (red).

Of course, there are always limitations when fitting a simple model to complex data, and these are discussed in detail in previous papers (see, e.g., Grier et al. 2017; Raimundo et al. 2020; Williams et al. 2020). Despite these limitations, the model is able to reproduce the observed emission-line shape and variability to the level of precision shown in the following section. The statistical temperature \( T \) accounts for the simplified nature of the BLR model as well as any fluctuations around the Gauss–Hermite fits to the emission line. The effect of \( T \) is reflected in all parameter uncertainties reported below.

3. Results

The modeling code is able to find regions of the BLR model parameter space that reproduce the observed emission-line shape and fluctuations (Figure 1). To avoid overfitting, we soften the likelihood function with a temperature \( T = 180 \) when constructing the posterior sample from the DNEST4 output, selected to be the lowest temperature for which signs of overfitting in the diagnostic plots are avoided. This is equivalent to multiplying all uncertainties on the spectra by a factor of \( \sqrt{T} = 13.4 \), which is reflected in the error bars in panels 4 and 5. The model displayed in Figure 1 was chosen to be representative of the full posterior sample and has parameter values close to those reported in Section 3.1. The large-scale variations in the C IV emission-line flux are very well captured by this model, although some of the shorter-timescale fluctuations (<100 days) are smoothed out. The structure that is visible in the normalized residuals corresponds to these shorter-timescale fluctuations, but we do not expect the smooth BLR model to fit all of these details.

3.1. BLR Model Parameters

Examining the posterior probability density functions (PDFs) for the model parameters, we find two clusters of solutions: those with \( \theta_i \) and \( \theta_o \) around 45° and those with \( \theta_i \) and \( \theta_o \) around 85°. Due to known degeneracies resulting from the flexible parameterization of the model (see, e.g., Grier et al. 2017), different combinations of parameters can produce identical particle distributions and velocities. For \( \theta_i \sim \theta_o \sim 85° \), the particles are arranged in a sphere, but as \( \gamma \sim 5 \), they are increasingly concentrated along the axis of the sphere, which is close to perpendicular to the observer’s line of sight. Since RM data cannot resolve rotations in the plane of the sky, these solutions are equivalent to near-face-on thick disk models and are thus degenerate with the first family of solutions. The 2D posterior PDFs reveal that this is the case for the models in our sample with \( \theta_i, \theta_o \sim 85° \). Unfortunately, the other model parameters are difficult to interpret in this arrangement, so we choose to exclude solutions with \( \theta_i > 65° \).

Taking the median and 68% confidence intervals for each parameter, we find a thick disk BLR with \( \theta_o = 52^{+6}_{-5} \) that is inclined by \( \theta_i = 40^{+8}_{-5} \). The parameter determining if emission is concentrated on the faces of the disk is not constrained for these models (\( \gamma = 2.6^{+1.4}_{-1.2} \)). The radial distribution of BLR emission drops off with radius faster than exponentially (\( \beta = 1.32^{+0.11}_{-0.09} \)), is shifted from the origin by a minimum radius \( r_{\text{min}} = 8.2^{+1.3}_{-1.1} \) ld, and has a median radius \( r_{\text{median}} = 33.0^{+5.4}_{-3.9} \) ld. This is visible in Figure 2 with the shell-like concentration of points around the BLR center and a rapid drop-off in point density at larger radii. The particles preferentially emit back toward the ionizing source (\( \kappa < -0.47 \)), which agrees with photoionization predictions, and the disk midplane is found to be mostly transparent (\( \zeta > 0.77 \)).

Kinetically, models are preferred in which most particles (\( f_{\text{clip}} = 0.83^{+0.04}_{-0.06} \)) are on near-circular orbits, which is reflected in the symmetric nature of the transfer function (Figure 3). The remaining particles have infalling trajectories (\( f_{\text{flow}} = 0.26^{+0.18}_{-0.13} \), \( \theta_e = 20.2^{+13.2}_{-13.2} \)). Finally, the black hole mass is very well
determined at log₁₀(M₆/₇9) = 8.31 ± 0.10. For comparison, Paper I measured log₁₀(M₆/₇9) = 8.63 ± 0.27 using the lag from cross-correlation paired with a scale factor f (Equation (1)) converted from H₃β-based measurements. The modeling approach is more reliable since it does not depend on H₃β-to-CIV conversions for f, and is more precise since it avoids the additional uncertainty introduced by the intrinsic scatter in f.

The rest-frame time lag one would measure based on these models is t₁_{median} = 36.4 ± 1.8 days, which agrees very well with the cross-correlation measurement of τ_{cen} = 36.5 ± 2.9 by Paper I. For the BLR model shown in Figure 3, we also compute the mean emission-line lag in the five wavelength windows used to measure the velocity-resolved lags in Paper I. We should note that the lag for the central wavelength bin in Paper I was based on a narrowband filter transmission curve, rather than the top-hat function used here. Since the transmission is higher near the center of the bin where the lags are longest, this will bias the filter-based measurements toward longer lags. Regardless, all of our model-based measurements are consistent with the velocity-resolved lags based on cross-correlation.

In Figure 4, we show the 2D posteriors for log₁₀(M₆/₇9) and t₁ and t₀. There is an anticorrelation between the black hole mass and inclination angle, which is expected—as the BLR is tilted closer to face on, a larger lag is required to reproduce the observed line width. We also find two streaks of solutions that can be separated by the line t₁ = t₀. Those with ϑᵢ > ϑ₀ tend to have slightly more transparent midplanes and a higher fraction of particles on near-circular orbits, but all other parameter distributions remain indistinguishable. Importantly, both cases give the same black hole mass.

### 3.2. Scale Factor f

Using the M₆/₇9 posterior PDF from our BLR model along with the CIV line widths and lags measured in Paper I, we can compute the appropriate scale factor, f, for the SDSS J2222+2745 BLR. We use the rest-frame lag, τ_{cen} = 36.5 ± 2.9 days, and the four emission-line widths: ΔV_{mean,FWHM} = 7734 ± 59 km s⁻¹, ΔV_{mean,s} = 4261 ± 49 km s⁻¹, ΔV_{max,FWHM} = 9219 ± 458 km s⁻¹, and ΔV_{max,s} = 5907 ± 148 km s⁻¹.

We propagate all uncertainties utilizing the full M₆/₇9 posterior as follows. For each sample in the posterior, we draw an emission-line width from a normal distribution, N(ΔV, σ₅²/V), where ΔV is the median value reported above and σ₅²/V is the corresponding scale factor. We then draw a time lag using the same approach, assuming an uncertainty that is equal to the average of the upper and lower uncertainties, 3.4 days. From this, we construct a sample of log₁₀(f) values and compute the median and 68% confidence intervals. We find log₁₀(f_{mean,σ}) = 0.20 ± 0.09 (f_{mean,σ} = 1.58 ± 0.32), log₁₀(f_{mean,FWHM}) = -0.32 ± 0.09 (f_{mean,FWHM} = 0.48 ± 0.11), log₁₀(f_{max,σ}) = -0.08 ± 0.07 (f_{max,σ} = 0.83 ± 0.12), and log₁₀(f_{max,FWHM}) = -0.46 ± 0.09 (f_{max,FWHM} = 0.35 ± 0.08). The uncertainties on log₁₀(f) are larger than those for log₁₀(M₆/₇9) since they include the uncertainties on ΔV and τ. An average virial
factor has been computed for the H\(\beta\) BLR by aligning the sample of reverberation-mapped AGNs with the \(M_{\text{BH}}-\sigma_{*}\) relation for inactive galaxies (e.g., Grier et al. 2013; Ho & Kim 2014), but no such calibrations exist for the CIV BLR.

Since analyses such as this one are both computationally and observationally expensive to carry out, we would like to be able to estimate \(f\) for individual AGNs based on other AGN properties. Williams et al. (2018) explored possible correlations between \(f\) and various BLR properties and AGN observables using the H\(\beta\) BLR. In Figure 5, we show \(f_{\text{CIV}}\) for SDSS J2222+2745 plotted against some of these parameters, along with the data and fits for \(f_{\text{H}\beta}\) by Williams et al. (2018) and \(f_{\text{CIV}}\) and \(f_{\text{H}\beta}\) for NGC 5548 (Williams et al. 2020). To determine the bolometric luminosity, we use \(\log_{10}(L_{1500}/\text{erg s}^{-1}) = 44.66 \pm 0.18\) from Paper I and assume a bolometric correction of \(BC_{1500} = 3.81\), computed by Shen et al. (2011) using the composite quasar spectral energy distributions of Richards et al. (2006). The H\(\beta\) and CIV-based scale factors are not necessarily expected to be the same due to differences in the two BLRs, so Paper I computed conversion factors between \(f_{\text{H}\beta}\) and \(f_{\text{CIV}}\) by comparing AGNs with line width and lag measurements available for both H\(\beta\) and CIV. They found that for a relation of the form \(\log_{10}(f_{\text{CIV}}) = \log_{10}(f_{\text{H}\beta}) + \alpha\), the best fits are \(\alpha_{\text{mean},\sigma} = 0.087 \pm 0.007\), \(\sigma_{\text{rms},\sigma} = -0.021 \pm 0.028\), and \(\alpha_{\text{mean,FWHM}} = 0.694 \pm 0.008\). In Figure 5, we also show the two CIV-based contours shifted by these conversion factors.

An anticorrelation between \(f\) and \(\theta_{i}\) is expected to exist, regardless of the emission line or AGN properties. The parameter \(\theta_{i}\) simply describes the orientation of the BLR in the sky, so for a BLR model with disk-like kinematics, as \(\theta_{i}\) increases, \(f\) must decrease to account for the larger line-of-sight velocity component contributing to the emission-line width. The SDSS J2222+2745 measurement happens to fall on the same relation as the H\(\beta\) BLR measurements, but it is possible that CIV BLRs follow a different relation and this is a chance alignment. A larger sample of CIV BLR models is necessary to distinguish between the two options.

We also find that the SDSS J2222+2745 measurement lies below the H\(\beta\)-based correlations in the \(\log_{10}(f)\) vs. \(\log_{10}(M_{\text{BH}}/M_{\odot})\) panels, and the offset is exaggerated by the \(f_{\text{CIV}}\) to \(f_{\text{H}\beta}\) correction factor. The H\(\beta\) relation between \(f_{\text{mean,FWHM}}\) and \(M_{\text{BH}}\) is currently detected only at the 2\(\sigma\) level (Williams et al. 2018), so a larger sample will be necessary to solidify the relationship (L. Villaflaena et al. 2021, in preparation). Assuming that the relationship is real for the H\(\beta\) BLR, though, the offset would represent a significant deviation. This suggests that either: (1) the correlations found by Williams et al. (2018) only hold for local AGNs and evolve as a function of redshift, or (2) the relation between \(f_{\text{H}\beta}\) and \(f_{\text{CIV}}\) is not as simple as that found by Paper I.

The first option might be the case if gas properties in BLRs changed between \(z \approx 2.8\) and \(z < 0.1\), affecting the BLR emissivity distribution. Unfortunately, SDSS J2222+2745 is the only AGN at high-\(z\) with modeling, so additional data is needed to explore this possibility. The second option is intriguing given the limited sample (six AGNs) from which the Paper I conversions were based. In fact, the offset between the Williams et al. (2020) \(f_{\text{H}\beta}\) point (solid black) and the converted \(f_{\text{CIV}}\) point (orange dashed) in Figure 5 demonstrates that the conversion was not appropriate for NGC 5548. The proper conversion likely depends on the BLR gas and AGN luminosity and would vary between objects.

BLR modeling of both the optical and UV BLRs for a single object can help uncover the differences between the BLRs that lead to the differences in scale factors. This has already been done for NGC 5548 (Williams et al. 2020) and a similar analysis will be performed for Mrk 817 using data from the AGN STORM 2 campaign that is currently ongoing (Kara et al. 2021). Photonization models can also be used to predict how the optical and UV BLRs will differ based on gas and ionizing flux conditions, and these are being implemented in the BLR model used in this work (P. R. Williams et al. 2021, in preparation).

4. Conclusions

The results presented here are the first of their kind for an AGN at \(z > 0.1\) and at the peak of AGN activity. Virtually all other \(M_{\text{BH}}\) measurements at high redshift are based on extrapolations from local measurements and uncertain conversions between the optical and UV BLR. Detecting signals such as the evolution of BH-host galaxy scaling relations requires a firmer understanding of the UV BLR and how it changes over time. The tight constraints on \(M_{\text{BH}}\) and other BLR properties for this object demonstrate that gravitational lensing is a powerful tool that can provide data good enough for detailed BLR studies.

While opportunities like that of SDSS J2222+2745 are rare, they serve an important role in expanding our understanding of the BLR outside of the local universe. With rapid improvements in lens detection techniques and the advent of the Vera C. Rubin Observatory Legacy Survey of Space and Time (LSST), more large separation lenses are expected to be found over the next decade (Oguri & Marshall 2010). Furthermore, the high cadence of LSST will enable monitoring of such systems to search for high-amplitude variability events that would trigger follow-up campaigns such as the one described in this paper. Estimating the expected number of lenses feasible for...
for this kind of analysis is beyond the scope of this paper, and is left for future work.

The main results of our analysis can be summarized as follows:

1. The C IV-emitting BLR for SDSS J2222+2745 is a thick disk with a size $r_{\text{median}} = 33.0_{-2.1}^{+2.4}$ light days. The kinematics are dominated by near-circular Keplerian motion, with the remaining 10%-20% of emission indicating infalling trajectories.

2. The median rest-frame lag produced by the BLR models is $36.4_{-1.8}^{+1.8}$ days which agrees closely with the $\tau_{\text{cen}} = 36.5_{-0.9}^{+2.9}$ day measurement from cross-correlation. The velocity-binned mean lags from the model are also consistent with the velocity-resolved lags from cross-correlation.

3. The black hole mass for SDSS J2222+2745 is $\log_{10}(M_{\text{BH}}/M_{\odot}) = 8.3_{-0.7}^{+0.6}$. This corresponds to scale factors of $\log_{10}(f_{\text{mean},c}) = 0.20_{-0.07}^{+0.09}$, $\log_{10}(f_{\text{mean},\text{FWHM}}) = -0.32_{-0.07}^{+0.09}$, $\log_{10}(f_{\text{rms},c}) = -0.08_{-0.07}^{+0.08}$, and $\log_{10}(f_{\text{rms},\text{FWHM}}) = -0.46_{-0.09}^{+0.10}$.

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Software: Astropy (Astropy Collaboration et al. 2013, 2018), DNEST4 (Brewer & Foreman-Mackey 2018), IRAF (Tody 1986, 1993), Scipy (Virtanen et al. 2020).

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References

Astropy Collaboration, Price-Whelan, A. M., Sipocz, B. M., et al. 2018, aj, 156, 123
Astropy Collaboration, Robitaille, T. P., Tollerud, E. J., et al. 2013, A&A, 558, A33
Blandford, R. D., & McKee, C. F. 1982, ApJ, 255, 419
Brewer, B., & Foreman-Mackey, D. 2018, Journal of Statistical Software, 86, 1
Brewer, B. J., Treu, T., Pancoast, A., et al. 2011, ApJL, 733, L33
Dahle, H., Gladders, M. D., Sharon, K., et al. 2013, ApJ, 773, 146
Ding, X., Silverman, J., Treu, T., et al. 2020, ApJ, 888, 37
Dyland, K. 2019, Master’s Thesis, Univ. Oslo https://www.duo.uio.no/handle/10852/69986?show=full
Ferrarese, L., & Ford, H. 2005, SSRv, 116, 523
Ferrarese, L., & Merritt, D. 2000, ApJL, 539, L9
Gebhardt, K., Bender, R., Bower, G., et al. 2000, ApJL, 539, L13
Grier, C. J., Martini, P., Watson, L. C., et al. 2013, ApJ, 773, 90
Grier, C. J., Pancoast, A., Barth, A. J., et al. 2017, ApJ, 849, 146
Ho, L. C., & Kim, M. 2014, ApJ, 789, 17
Hook, I. M., Jørgensen, I., Allington-Smith, J. R., et al. 2004, PASP, 116, 425
Kara, E., Mehdipour, M., Kriss, G. A., et al. 2021, arXiv:2105.05840
Karo, E., Mehdipour, M., Kriss, G. A., et al. 2021, arXiv:2105.05840
Oguri, M., & Marshall, P. J. 2010, MNRAS, 405, 2579
Pancoast, A., Brewer, B. J., & Treu, T. 2011, ApJL, 730, 139
Pancoast, A., Brewer, B. J., Treu, T., et al. 2012, ApJL, 754, 49
Pancoast, A., Brewer, B. J., Treu, T., et al. 2014, MNRAS, 445, 3073
Park, D., Kelly, B. C., Woo, J.-H., & Treu, T. 2012, ApJS, 203, 6
Peterson, B. M. 1993, PASP, 105, 247
Peterson, B. M. 2014, SSRv, 183, 253
Raimundo, S. I., Vestergaard, M., Goad, M. R., et al. 2020, MNRAS, 493, 1227
Richards, G. T., Lacy, M., Storrie-Lombardi, L. J., et al. 2006, ApJS, 166, 470
Shen, Y., Richards, G. T., Strauss, M. A., et al. 2011, ApJS, 194, 45
Tody, D. 1986, Proc. SPIE, 627 ed. D. L. Crawford., 733
Tody, D. 1993, in ASP Conf. Ser. 52, Astronomical Data Analysis Software and Systems II, ed. R. J. Hanisch, R. J. V. Brissenden, & J Barnes (San Francisco, CA: ASP), 173
Virtanen, P., Gommers, R., Oliphant, T. E., et al. 2020, Nature Methods, 17, 261
Williams, P. R., Pancoast, A., Treu, T., et al. 2018, ApJ, 866, 75
Williams, P. R., Pancoast, A., Treu, T., et al. 2020, ApJ, 902, 74
Williams, P. R., Treu, T., Dahle, H., et al. 2021, ApJ, 911, 64

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