



AGN STORM 2. II. Ultraviolet Observations of Mrk 817 with the Cosmic Origins Spectrograph on the Hubble Space Telescope*

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Abstract

We present reverberation mapping measurements for the prominent ultraviolet broad emission lines of the active galactic nucleus Mrk 817 using 165 spectra obtained with the Cosmic Origins Spectrograph on the Hubble Space Telescope. Our ultraviolet observations are accompanied by X-ray, optical, and near-infrared observations as part of the AGN Space Telescope and Optical Reverberation Mapping Program 2 (AGN STORM 2). Using the cross-correlation lag analysis method, we find significant correlated variations in the continuum and emission-line light curves. We measure rest-frame delayed responses between the far-ultraviolet continuum at 1180 Å and Ly α λ 1215 Å ($10.4^{+1.6}_{-1.4}$ days), N V λ 1240 Å ($15.5^{+1.0}_{-4.8}$ days), Si IV +]O IV λ 1397 Å ($8.2^{+1.4}_{-1.4}$ days), C IV λ 1549 Å ($11.8^{+3.0}_{-2.8}$ days), and He II λ 1640 Å ($9.0^{+4.5}_{-1.9}$ days) using segments of the emission-line profile that are unaffected by absorption and blending, which results in sampling different velocity ranges for each line. However, we find that the emission-line responses to continuum variations are more complex than a simple smoothed, shifted, and scaled version of the continuum light curve. We also measure velocity-resolved lags for the Ly α and C IV emission lines. The lag profile in the blue wing of Ly α is consistent with virial motion, with longer lags dominating at lower velocities, and shorter lags at higher velocities. The C IV lag profile shows the signature of a thick rotating disk, with the shortest lags in the wings, local peaks at ± 1500 km s $^{-1}$, and a local minimum at the line center. The other emission lines are dominated by broad absorption lines and blending with adjacent emission lines. These require detailed models, and will be presented in future work.

Unified Astronomy Thesaurus concepts: Active galactic nuclei (16); Reverberation mapping (2019); Active galaxies (17); Seyfert galaxies (1447)

Supporting material: machine-readable tables

1. Introduction

Supermassive black holes (SMBHs) are among the most extreme objects in the universe, and rapidly grow during an active phase of accretion. The symbiotic correlation between SMBH mass (M_{BH}) and galaxy bulge properties (Magorrian et al. 1998; Ferrarese & Merritt 2000; Gebhardt et al. 2000; Gültekin et al. 2009; Kormendy & Ho 2013) implies that black holes are essential ingredients in understanding galaxy evolution. During the active phase of accretion, the galaxy fuels the black hole while outflows in the form of winds and jets may regulate black hole–galaxy coevolution by removing gas from the host galaxy and shutting down (Di Matteo et al. 2005; Hopkins et al. 2008) or enhancing star formation (Chambers et al. 1990; Nesvadba et al. 2020). Despite decades of research, the cosmic evolution of SMBHs and their role in galaxy formation and evolution is still not fully understood. Revealing the nature of gas outflows and their origin near the central black hole are among the foremost requirements for understanding the growth history of SMBHs over cosmic time.

The masses of nearby SMBHs are generally measured using high spatial resolution observations of gas or stellar dynamics (for a review, see Kormendy & Ho 2013) with a few exceptions. The mass of Sagittarius A* has been measured using astrometric monitoring of the orbits of individual stars in the central few parsecs of the Milky Way (Genzel et al. 2000; Ghez et al. 2000). More recently, the mass of Sagittarius A* and the SMBH in M87 have been measured from radio observations of their black hole *shadow* (Event Horizon

Telescope Collaboration et al. 2019; Akiyama et al. 2022). There are also interferometric observations of 3C273 (Gravity Collaboration et al. 2018; Wang et al. 2020; Li et al. 2022) and NGC 3783 (Gravity Collaboration et al. 2021). Most of these measurements are not possible for more distant galaxies (>100 Mpc) even with next-generation facilities.

Over the past few decades, reverberation mapping (RM; Blandford & McKee 1982; Peterson 1993) has emerged as a powerful technique to study the geometry and kinematics of the gas surrounding SMBHs, and as a tool to estimate the mass of the BHs (for a recent review of multiscale RM, see Cackett et al. 2021). Nearly all rapidly accreting SMBHs, observed as broad emission-line active galactic nuclei (AGN), exhibit variability on timescales of weeks to years (e.g., Collier et al. 2001; Peterson et al. 2004; Kelly et al. 2009; MacLeod et al. 2012). The gas falling onto the central black hole forms an accretion disk that emits UV photons and ionizes a broad emission-line region (BLR) composed of high-density, high-velocity gas. In its most simplistic implementation, RM can be used to measure the mean light-travel time across the BLR as the time delay (or lag), τ , between the variability in the continuum and the subsequent response of the broad emission-line gas. Assuming that this gas is dominated by the gravitational field of the central black hole, the broad emission-line width (as a line-of-sight gas velocity) is combined with the responsivity-weighted BLR size to obtain an M_{BH} estimate (Peterson & Wandel 2000).

However, the RM M_{BH} estimate relies on a dimensionless factor f , of order unity, that depends on the geometry of the BLR and its orientation relative to the observer's line of sight (Onken et al. 2004; Woo et al. 2010; Park et al. 2012; Grier et al. 2013), which remains poorly understood despite decades of research. Hydrodynamical modeling shows that disk winds can further complicate the measurement of f (Kashi et al. 2013). Some studies of BLR structure support a disk-like BLR (Wills & Browne 1986; Eracleous & Halpern 1994;

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Vestergaard et al. 2000; Eracleous & Halpern 2003; Strateva et al. 2003; Jarvis & McLure 2006; Gezari et al. 2007; Lewis et al. 2010; Storchi-Bergmann et al. 2017). Other studies suggest that radiation pressure significantly contributes to the BLR dynamics (Marconi et al. 2008; Netzer & Marziani 2010). Further ambiguity in the dynamics of the BLR is caused by the mixed evidence for inflow and outflow processes on these spatial scales (e.g., Bentz et al. 2009; Denney et al. 2009; Barth et al. 2011a, 2011b; Du et al. 2016; Pei et al. 2017).

A more powerful application of RM is the measurement of velocity-resolved responses of the broad emission lines to continuum flux variations (Bahcall et al. 1972; Blandford & McKee 1982). Velocity-resolved RM constrains both the distance and kinematics of the broad-line-emitting gas, and enables detailed dynamical modeling of the BLR, but it requires much higher-quality data than velocity-integrated RM (Horne et al. 2004). These measurements recover a projection of the BLR structure and kinematics into two observables, the line-of-sight velocity and the emission-line time lag. The distribution of lags can be described as a function of velocity, and it is referred to as the *velocity-delay map*. The broad emission-line flux variations, $\Delta L(V, t)$, are expressed by the convolution of the continuum flux variations $\Delta C(t)$ with the velocity-delay map, $\Psi(V, \tau)$ (also known as the response function) as

$$\Delta L(V, t) = \int_0^\infty \Psi(V, \tau) \Delta C(t - \tau) d\tau. \quad (1)$$

The primary goal of velocity-resolved RM is to directly constrain the BLR geometry and kinematics either through the inverse problem approach (by reconstructing the velocity-delay map) or through forward modeling. Lag asymmetries between the blue and red wings of a line can be an indication for nonvirial motion of the BLR gas.

Even though these velocity-resolved lags enable simple and qualitative inferences about the BLR gas dynamics, detailed interpretations are dependent on orientation, geometrical complexities, and assumptions about the optical depth in the emission-line region (see, e.g., Welsh & Horne 1991; Goad et al. 2012). In general, inflows cause shorter lags in the red wing, while outflows lead to shorter lags in the blue wing. High-quality velocity-resolved RM campaigns have been successful for a handful of nearby AGN using ground-based observations (Bentz et al. 2008, 2009; Denney et al. 2009; Bentz et al. 2010; Grier et al. 2013; Du et al. 2016; De Rosa et al. 2018; Du et al. 2018; Brotherton et al. 2020; Bentz et al. 2021; Bao et al. 2022; U et al. 2022). These observations have provided results that are consistent with gas in elliptical orbits for some objects, while others indicate either inflowing or outflowing gas trajectories. Furthermore, the kinematics of the H β -emitting region are not uniform across all AGN, with different objects showing a mix of inflowing, stable, and outflowing gas dynamics (Pancoast et al. 2014; Li et al. 2018; Williams et al. 2018; U et al. 2022; Villafañã et al. 2022). Velocity-resolved RM observations for higher ionization ultraviolet (UV) emission lines were only achieved recently through the AGN Space Telescope Optical Reverberation Mapping (AGN STORM) program.

The first AGN STORM campaign was a pioneering multi-wavelength reverberation mapping program, targeting NGC 5548 over 6 months with daily spectroscopic observations using the Hubble Space Telescope (HST) Cosmic Origins

Spectrograph (COS) in 2014 (De Rosa et al. 2015). The HST observations were also supported by coordinated observations from Swift using the XRT and UVOT instruments (Edelson et al. 2015), intensive ground-based photometric and spectroscopic monitoring (Fausnaugh et al. 2016; Pei et al. 2017), and four Chandra X-ray observations distributed over the duration of the campaign (Mathur et al. 2017). Key results from the AGN STORM campaign include the following.

1. Identification of a stratified BLR with mean broad emission-line delays spanning 2.5–7 days with respect to the $\lambda 1367$ Å continuum, depending on the emission line (De Rosa et al. 2015). When placed in the context of previous intensive monitoring campaigns of this source (e.g., Clavel et al. 1991; Peterson et al. 1991; Denney et al. 2009), the continuum fluctuation timescales were substantially shorter even though continuum fluctuation amplitudes and average luminosities were similar. The shorter emission-line delays and smaller variability amplitudes were in part due to the short timescale continuum fluctuations.
2. The relative time lags between the variations in the inner/shorter wavelengths and outer/longer wavelengths of the disk were measured through continuum reverberation from the UV through the near-infrared (NIR) wavelength region. These measurements are broadly consistent with the temperature gradient expected from the foundational *thin-disk* model of Shakura & Sunyaev (1973), but suggest a disk that is 3 times larger (McHardy et al. 2014; Edelson et al. 2015; Fausnaugh et al. 2016). However, these measurements may be affected by contributions from diffuse continuum emission from the BLR clouds (Korista & Goad 2001; Chelouche et al. 2019; Korista & Goad 2019).
3. The appearance of anomalous BLR behavior (the so-called *BLR holiday*) started midway through the campaign (Goad et al. 2016; Pei et al. 2017; Dehghanian et al. 2019a; Kriss et al. 2019). This was a 40 day period where the emission-line light curve was uncorrelated with the continuum. Anomalous behavior was also later identified in the continuum bands (Goad et al. 2019), further supporting contamination of the UV-optical continuum by the diffuse continuum from the BLR.
4. Recovery of the most detailed velocity-delay maps ever obtained for the prominent emission lines revealed kinematics that are dominated by near-circular Keplerian motion, and a weaker response from the far side of a somewhat flattened BLR geometry (Horne et al. 2021).
5. The black hole mass derived from a reverberation analysis using a mean time delay, a characteristic emission-line velocity dispersion, and a scale factor of $\langle f \rangle \approx 5$ is in good agreement given the uncertainties with the mass derived from dynamical modeling utilizing velocity-delay mapping information (Williams et al. 2020). The masses inferred from independent emission lines are mutually consistent, even though the geometry and kinematics of each line-emitting region are different.

The present work describes the results of a similar RM campaign on Mrk 817, a large monitoring program built around 165 epochs of HST observations.⁵² Hereafter, we refer to the campaign on NGC 5548 as AGN STORM 1 and the current program on Mrk 817 as AGN STORM 2.

⁵² HST-GO-16196; Peterson et al. (2020).

The target for AGN STORM 2, Mrk 817, was carefully selected because it showed no history of strong UV absorption and/or X-ray obscuration. Therefore, it was less likely to be obscured, which was a major source of complications in AGN STORM 1 (Goad et al. 2016; Dehghanian et al. 2019a; Kriss et al. 2019). Historically, Mrk 817 has exhibited weak, variable intrinsic narrow absorption features, which can be used to probe the unobservable UV ionizing continuum along the line of sight (Dehghanian et al. 2019a, 2019b; Kriss et al. 2019). The Mrk 817 SMBH has similar mass to that of NGC 5548 ($M_{\text{BH}} \approx 3.85 \times 10^7 M_{\odot}$); however, its higher Eddington ratio ($L/L_{\text{Edd}} \approx 0.2$ compared to $L/L_{\text{Edd}} \approx 0.03$ for NGC 5548) probes a new region of parameter space compared to NGC 5548. Furthermore, the higher redshift of Mrk 817 ($z = 0.03146$) ensures that the $\text{Ly}\alpha$ line is not affected by geocoronal emission and Galactic absorption features. Also, Mrk 817 is located near the ecliptic pole, making it continuously observable.

Even though Mrk 817 was carefully selected to be free of strong UV absorption lines, the AGN STORM 2 observations reveal the presence of new broad and narrow absorption lines, which may indicate the appearance of a new dust-free wind located at the inner BLR that partially obscures the line of sight to the SMBH (Kara et al. 2021). Furthermore, coordinated X-ray observations with XMM-Newton and NICER show that the X-rays are significantly obscured compared to earlier observations (see Miller et al. 2021 for an independent analysis using NuSTAR data). Kara et al. (2021) provide a detailed discussion of these early results, describing the HST program from the first 97 days of the AGN STORM 2 campaign.

This paper is the second in a series of the AGN STORM 2 campaign. We discuss the calibration and improvements of the spectra from the HST Cosmic Origins Spectrograph (COS; Green et al. 2012) for the full duration of the AGN STORM 2 campaign. In Section 2 we discuss the details of the observations. Section 3 describes our custom reduction pipeline. We present the analysis of spectra, flux, and light-curve measurements in Section 4. In Section 5 we describe the light curves and their cross correlations. In Section 6 we discuss the velocity-binned reverberation lags for the major UV emission lines. Throughout this work, we adopt a Lambda cold dark matter cosmology with $\Omega_{\Lambda} = 0.7$, $\Omega_M = 0.3$, and $H_0 = 70 \text{ km s}^{-1} \text{ Mpc}^{-1}$.

2. Observations

2.1. Program Design

The ultimate goal of the AGN STORM 2 campaign is to understand the origins of gas flows in the nuclear region, including the location and dynamics of the BLR. This is achieved through the construction of velocity-delay maps or through forward modeling using high-quality RM observations. To obtain such a data set, several considerations are required in the design of the observations, as follows.

1. Observing cadence: A velocity-delay map with time resolution Δt requires data with time sampling on scales $\Delta t/2$ or shorter.
2. Signal-to-noise ratio (S/N) and spectroscopic calibration: Continuum and emission-line variations are small on short timescales and therefore high S/N is required to successfully detect them. In practice, this means

$S/N \geq 100$ in the integrated continuum windows (measured over several resolution elements), and local flux calibrations precise and stable to $\sim 1\%$ – 2% .

3. Spectral resolution: The line-of-sight (LOS) velocity resolution should be sufficient to resolve structures in the emission and absorption-line profiles on scales of a few hundred kilometers per second.
4. Duration: The key to obtaining a clear reverberation signal is structure in the light curves. The temporal coverage must be long enough to both have a significant probability of including such features and track the subsequent response of the BLR.

Previously, Mrk 817 has been the target of ground-based RM campaigns (Peterson et al. 1998; Denney et al. 2010). These campaigns show an optical continuum rms variability amplitude of 5%–14% on timescales of weeks to months. The AGN UV variability is expected to be twice as large (Korista et al. 1995; Marshall et al. 1997; MacLeod et al. 2012). The $\text{H}\beta$ BLR size in Mrk 817 (14–34 light days) is larger compared to the BLR size in NGC 5548, which was observed during the AGN STORM 1 campaign. For a historically variable target such as Mrk 817, an every-other-day observing cadence over 365 days provides the necessary duration while also capturing UV-emitting-region variability (compared to daily cadence over a 6 month duration in the AGN STORM 1 campaign).

2.2. HST Observations

Mrk 817 was observed with HST COS over 165 epochs with a median cadence of 2 days from 2020 November 24 to 2022 February 24. These data are available from MAST: doi:10.17909/10sp-zt74. We obtained high-resolution COS spectra in single-orbit visits using the G130M ($10,000 < R < 15,000$) and G160M ($13,000 < R < 24,000$) gratings to obtain far-UV (FUV) spectra covering 1070–1750 Å. For each visit, we obtained four 60 s exposures with the G130M grating centered at 1222 Å, two exposures of 175 and 180 s using the G160M grating centered at 1533 Å, and two 195 s exposures with G160M at 1577 Å. These exposure times ensure an $S/N \approx 10$ per resolution element in the continuum at 1180 and 1502 Å.

For COS FUV observations, moving the spectra slightly in the dispersion direction minimizes the effects of small-scale fixed-pattern noise, detector artifacts, and grid-wire shadows in the detector, and improves the S/N of the coadded spectra (Dashtamirova & Fischer 2020). Furthermore, prolonged exposure of the same detector area to bright targets causes it to become less efficient at photon-to-electron conversion over time (a phenomenon also known as *gain-sag*). While the bright $\text{Ly}\alpha$ airglow falls in the detector gap for the G130M/1222 setting, repeated illumination by the AGN emission lines highly affects the detector sensitivity. To mitigate the issue, we ensured that the location of the spectrum on the detector was periodically moved by alternating the G130M/1222 exposures between four grating offset focal-plane positions (FP-POS = 1–4), the G160M/1533 exposures between two FP-POS configurations (FP-POS = 1, 2), and the G160M/1577 exposures between the remaining 2 FP-POS settings (FP-POS = 3, 4).

We also used 13 additional orbits to observe a flux standard star (WD0308-565) periodically throughout the campaign, in order to obtain high-precision flux calibrations on small spectral scales, as outlined in Section 2.1. We observed

WD0308-565 approximately once per month, using exactly the same settings as the RM program.

In addition to the COS spectra, we dedicated six additional orbits to obtain low-resolution spectra using the STIS/NUV-MAMA and the STIS CCD to capture full spectra from the FUV to the near-IR. We used gratings G230L, G430L, and G750L in 2-orbit visits paired with COS observations to cover the range of 940–10,200 Å. These visits were evenly distributed throughout the campaign and they were coordinated with XMM-Newton observations, as described by Kara et al. (2021). The full-spectrum snapshots plus the X-ray spectra provide information on the spectral energy distribution of Mrk 817. The close agreement (generally <5%) between the 1600 and 1700 Å fluxes in the STIS observations (which used a 0".2 slit) and the COS observations in the same visit (2".5 circular aperture) show that any host galaxy contribution to the COS continuum flux is negligible. The results of the combined COS and STIS spectra will be released in a separate paper.

Out of 165 epochs, eight visits suffered a failed guide-star acquisition or a guide-star loss of lock, leading to empty exposures. Furthermore, HST suffered two safe-mode incidents resulting in large gaps in our campaign. The first HST safing incident occurred on 2021 June 12 and lasted for 36 days. The second HST safing occurred on 2021 October 25, at the beginning of an expected 16 day observing gap due to no guide-star availability, and lasted for 51 days. Our original observing plan anticipated that 10% of visits might fail due to HST gyro issues, which is similar to the final rate of failed visits (11%). All of the lost visits were compensated and added to the end of the campaign at nominal cadence. As a result, the observing campaign ended on 2022 February 24, instead of the expected end date of 2021 November.

3. Data Reduction

We used the CalCOS pipeline v3.4.1 for the bulk of our data processing. In order to successfully perform spectral modeling and velocity-resolved RM, we need the spectra to be locally⁵³ accurate, precise, and stable. The standard COS reduction pipeline guarantees a flux accuracy at the 5% level, a global precision of 2%, and a wavelength accuracy of better than 10 km s⁻¹ for G130M/G160M gratings. While the absolute flux accuracy we require is achieved with the automated reduction pipeline, we need the flux measurements to be locally precise to <2%. Therefore, after testing the repeatability and precision of the wavelength solution, we refined the existing flux calibration reference files and applied a post-CalCOS pipeline to further process the data, following a similar approach to the one described by De Rosa et al. (2015) in AGN STORM1. The main areas of improvement include refinements in the sensitivity function and the time-dependent sensitivity (TDS) correction, as outlined below. These reprocessed data and the various forms of combined spectra described below are available as a high-level science product in the Mikulski Archive for Space Telescopes (MAST)⁵⁴ as the data set identified by doi:10.17909/n734-k698. The reference files used in the custom pipeline are also available through the same DOI.

3.1. Dispersion Solution

To verify the precision of the dispersion solution for each separate grating setting and each separate observation, we compare the registration of all the campaign spectra. In absolute terms, we also compare the mean positions of selected interstellar medium (ISM) lines to direct line-of-sight observations of Galactic H I 21 cm emission from Murphy et al. (1996). The H I emission is slightly blueshifted by -16 km s⁻¹ in the heliocentric frame, with additional components extending blueward to -80 km s⁻¹. The ISM lines in the COS spectrum have a similar structure, with the deepest point in the troughs agreeing with the H I to better than 10 km s⁻¹.

3.2. Sensitivity Function and TDS

To obtain a local precision of the flux calibrations at better than 2%, we used approximately monthly observations of the CalSpec calibration standard WD0308-565. The standard-star observations were obtained using exactly the same gratings, central wavelength (CENWAVE) settings, and focal-plane positions (FP-POS) as the science exposures. We estimate the quality of the flux calibration by inspection of the fractional residuals

$$f_{\text{res}} = \frac{f_{\text{WD,Obs}} - f_{\text{Model}}}{f_{\text{Model}}}, \quad (2)$$

of the calibrated standard-star spectra and their respective CalSPEC stellar model (we used the latest model available for WD0308-565), where both $f_{\text{WD,obs}}$ and f_{Model} are binned over 1 Å using a boxcar filter in order to increase the S/N per spectral element. By analyzing these calibration data, we verified that the residuals at each epoch, as well as their variations through the campaign, were higher than the AGN STORM2 requirements and thus need further calibrations.

The COS flux calibration is done in two steps: (a) characterization of the time evolution of the sensitivity through the TDS correction, and (b) derivation of static sensitivity functions, obtained once for each lifetime position.

The COS reference TDS model is obtained from approximately bimonthly observations of the standard stars WD0308-565 and GD71 as part of the yearly COS TDS monitoring program. Observations are obtained for a selected subset of CENWAVES, including the bluest and reddest CENWAVE for each grating at one fixed FP-POS (FP-POS = 3). The COS monitoring program covers all of the CENWAVES used in our observations (1222 Å, 1533 Å, 1577 Å). However, for some of the CENWAVES (e.g., G160M/1577), the FUVB and FUVB sensitivity functions are obtained from observations of different CalSPEC standards, which can affect the resulting spectral shape due to intrinsic systematic uncertainties in the stellar models. To improve the COS monitoring time cadence in our custom reduction, we utilized the existing COS monitoring program data and additionally added our observations of WD0308-565 at FP-POS = 3 only for G130M/1222 and G160M/1577 and FP-POS = 1 and 2 for G160M/1533 because no COS monitoring data is available for FUVB with WD0308-565 for that CENWAVE. We then refined the TDS model by increasing the number of time intervals over which the TDS trends are computed and by redefining the wavelength ranges used in the analysis.

⁵³ Single-pixel level.

⁵⁴ <https://archive.stsci.edu/>

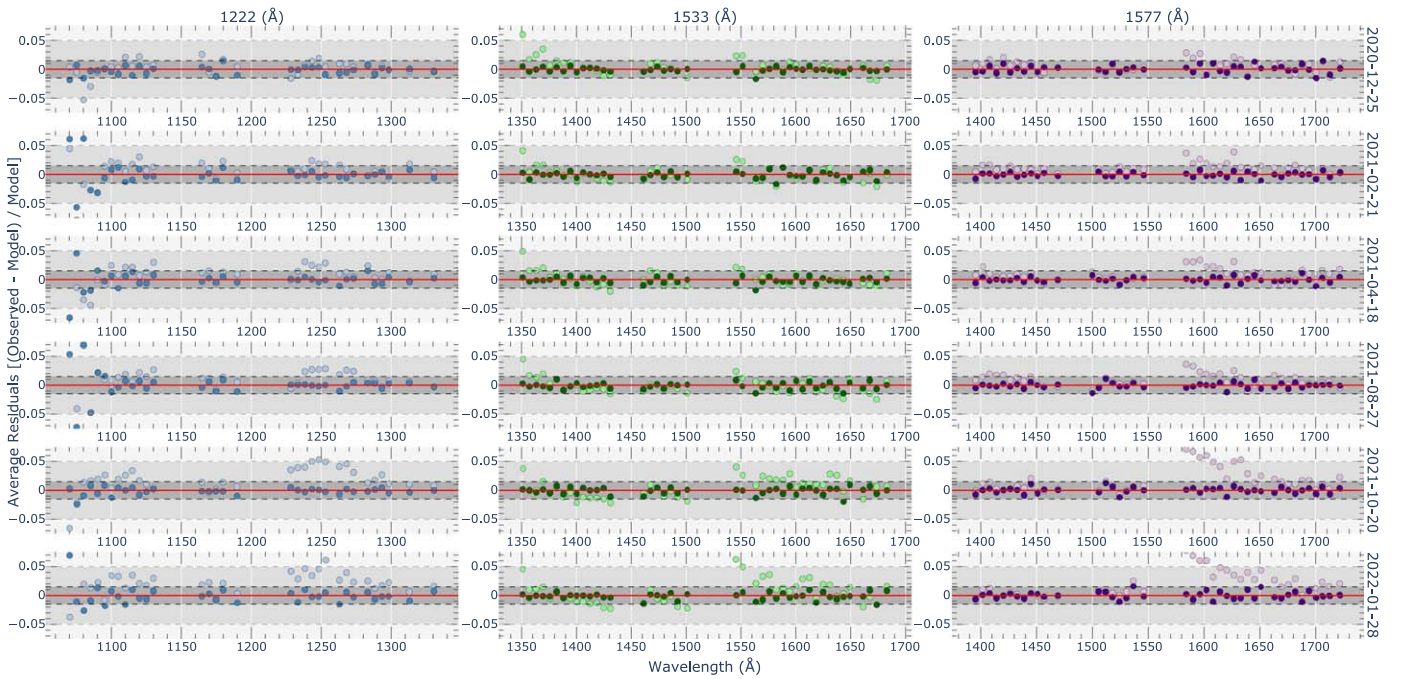


Figure 1. Flux residuals f_{res} vs. wavelength for each of the three central wavelengths used to observe Mrk 817: 1222 Å (left, blue), 1533 Å (middle, green), and 1577 Å (right, purple). Six of the twelve white-dwarf epochs are shown. In each panel, the lighter circles represent the residuals obtained with CalCOS, while the darker circles are obtained with the AGN STORM 2 updated data reduction. The light gray band and dark gray band show $\pm 5\%$ and $\pm 1.5\%$ residuals, respectively.

In the standard CalCOS reduction, one sensitivity function is used per lifetime position and any time variations are resolved in the TDS correction. However, we see variations in the fractional residuals (Equation (2)) over time that could not be accounted for with any additional TDS corrections. We therefore believe that the variations may come from uncharacterized changes in the standard flux or flats, so, to improve the flux calibrations, we obtained 12 new sensitivity functions using our monthly observations of WD0308-565. The Mrk 817 data were then reduced using the sensitivity function that was truncated in time to the science observation date and the updated TDS correction, where any observation before that time was reduced using the sensitivity function from the previous month. The improved residuals of our calibrations are illustrated in Figure 1.

We use the standard deviation of the residuals for each grating as an estimate of the fractional precision error δP to be associated with each epoch. δP is a proxy for the stability of the flux calibration at a given time. Since our final G160M spectra are obtained from the combination of two CENWAVEs, we conservatively define the fractional precision error for the G160M grating as the maximum fractional uncertainty computed for the 1533 and 1577 Å CENWAVEs, respectively. Table 1 lists the final fractional precision as a function of time for G130M and G160M.

3.3. Spectral Combination and Error Arrays

The final data product consists of one combined spectrum per grating per epoch. The final spectra are binned by 4 pixels in order to increase the S/N per spectral element of the AGN continuum. This binning still results in two binned macropixels per COS resolution element. For the G160M grating, the spectra from the G160M/1533 and G160M/1577 settings are merged together and resampled over a 4 pixel interval, using a python implementation of the iraf/splice algorithm. Each

Table 1
Fractional Precision Error δP

HJD Start (Days)	HJD End (Days)	G130M %	G160M %
9178	9206	0.8	0.6
9207	9229	0.8	0.8
9230	9264	0.8	0.7
9265	9296	0.9	0.8
9297	9320	0.5	0.8
9321	9411	0.8	0.6
9412	9451	1.3	0.7
9452	9476	0.9	0.7
9477	9505	0.9	0.7
9506	9547	0.8	0.7
9548	9605	1.6	0.7
9606	9633	0.9	0.7

pixel is weighted by the DQ_WGT array in the COS data files to correctly account for masked bad pixels. Finally, we modify the flux error obtained by propagating the CalCOS statistical uncertainty to include a correction for low number statistics following Gehrels (1986); the correction is applied in counts space. This is particularly important for regions of the combined spectra characterized by low counts (e.g., the blue end of G130M, the red end of G160M, and the troughs of the deepest absorption lines).

4. Data Analysis

To measure spectral variations and time-series quantities, we follow steps similar to those of De Rosa et al. (2015). We first compute the mean and root-mean square (rms) spectra from the 165 epochs of G130M and G160M observations. The rms spectra are used to isolate the variable spectral components. This clears the way for defining integration limits for the

Table 2
Integration Limits for Light Curves

Component	Integration Limit Å	Shortward Continuum Å	Longward Continuum Å	Total Line Flux Fraction ^a	Velocity Range km s ⁻¹
$F_{\lambda}(1180)$	1173.0–1185.0	...	...	...	...
$F_{\lambda}(1398)$	1394.2–1401.6	...	...	...	...
$F_{\lambda}(1502)$	1493.0–1511.0	...	...	...	...
$F_{\lambda}(1739)$	1736.0–1741.5	...	...	...	...
Ly α	1235.0–1248.0	1173.0–1185.0	1318.0–1330.0	21%	–1500 to –4500
N V	1268.0–1290.0	1173.0–1185.0	1318.0–1330.0	64%	–2250 to +2750
Si IV + O IV	1424.0–1460.0	1394.2–1401.6	1493.0–1511.0	77%	–4250 to +3000
C IV	1590.0–1638.0	1493.0–1511.0	1736.0–1741.5	63%	–1250 to +7500
He II	1680.0–1700.0	1493.0–1511.0	1736.0–1741.5	60%	–2000 to +1500

Note.

^a The total line flux fraction was obtained by integrating over the model of the mean spectrum for each emission line.

continuum and line light curves as outlined below (see Table 2).

4.1. Mean and rms Spectra

The mean spectrum for the set of G130M and G160M spectra is defined as

$$\bar{F}(\lambda) = \frac{1}{N} \sum_{i=1}^N F_i(\lambda) \quad (3)$$

where F_i is the i th spectrum of the series of $N=165$ spectra, and the rms spectrum is defined as

$$S(\lambda) = \left(\frac{1}{N-1} \sum_{i=1}^N (F_i(\lambda) - \bar{F}(\lambda))^2 \right)^{1/2}. \quad (4)$$

The mean and rms spectra for the G130M and G160M settings are shown in Figures 2 and 3, respectively. The statistical uncertainty in the mean spectra is

$$\sigma_{\bar{F}}(\lambda) = \frac{1}{N} \left[\sum_{i=1}^N \sigma_{F_i}^2(\lambda) \right]^{1/2}, \quad (5)$$

where σ_{F_i} is the error spectrum of the i th spectrum in the series. We estimate the total uncertainty of the mean spectrum as the quadratic sum of the statistical uncertainty and fractional precision δ_p .

The rms spectrum computed from Equation (4) and illustrated in blue in Figures 2 and 3 includes the intrinsic variability, σ_0 , as well as the variance due to noise (Park et al. 2012; Barth et al. 2015). We follow De Rosa et al. (2015) and model the distribution of the residuals of each pixel about the mean to estimate σ_0 using a maximum-likelihood estimator for the optimal average. The estimated intrinsic rms spectrum, σ_0 , obtained from the 165 HST COS spectra is shown in red (see Figures 2 and 3).

The intrinsic rms spectrum reveals several significant features of the spectral variability observed in the AGN STORM 2 campaign. The broad emission lines themselves, the primary motivation for the campaign, show strong variability across their entire profiles. However, the variability amplitude varies across the different line components. For example, in the He II line, the majority of the variability is in a very broad component. There is also a strong rise in variability at the blue end of the spectrum, which is due to the red wing of the broad O VI $\lambda 1033$ Å emission line. Furthermore, we note that the C IV

rms profile is broader compared to its mean profile, which is similar to other studies of the C IV emission line that could be indicative of a non-variable core component (Korista et al. 1995; Denney 2012), and has raised concerns over the suitability of C IV for SE M_{BH} estimates.

The broad absorption troughs also produce distinctive signatures. The saturated portions of the troughs typically show less variability than surrounding portions of the spectrum. This is particularly noticeable in Ly α , N V, and the centers of the C IV doublet. Both transitions in the N V doublet display prominent troughs at the peak of Ly α at 1254 Å and on its red wing at 1259 Å. Note that these saturated troughs have shapes quite similar to the saturated broad Ly α trough at 1230 Å. The weaker, potentially more optically thin portions of the obscuring outflow, however, show greater variability, perhaps indicating a more immediate response to continuum variations. This is evident in the C III* $\lambda 1176$ trough at 1189 Å, the troughs in the Si IV doublet, and the red and blue wings of the C IV trough.

We take the presence of these features into account when we evaluate the windows for extracting continuum and emission-line fluxes in the next section.

4.2. Integrated Light Curves

We extract the light curves for the continuum and the prominent UV emission lines listed in Table 2 by performing the spectral flux measurement in the observed frame. We have not corrected the spectra for Galactic extinction in order to facilitate the cleanest comparison with other measurements to be reported elsewhere in this series of papers (e.g., broadband photometry).

There are bad pixels throughout the spectrum, and their location and severity change with time and instrument settings. To prevent the introduction of artificial variations in the relative-flux estimates, bad pixels are masked throughout the data set. This means that if a pixel is bad in any of the visits, the pixel is masked out in all of the 165 spectra. We have also masked out Galactic Ly α absorption in the range 1209–1222 Å. The presence of intrinsic broad and narrow absorption features that appeared during the AGN STORM 2 campaign complicates the analysis of the spectra. Final integration ranges (listed in Table 2) were chosen using the empirical model of the spectra and the variability characteristics of the rms spectrum. Our modeling is similar to that used by Kriss et al. (2019) for the AGN STORM 1 campaign, which includes the broad

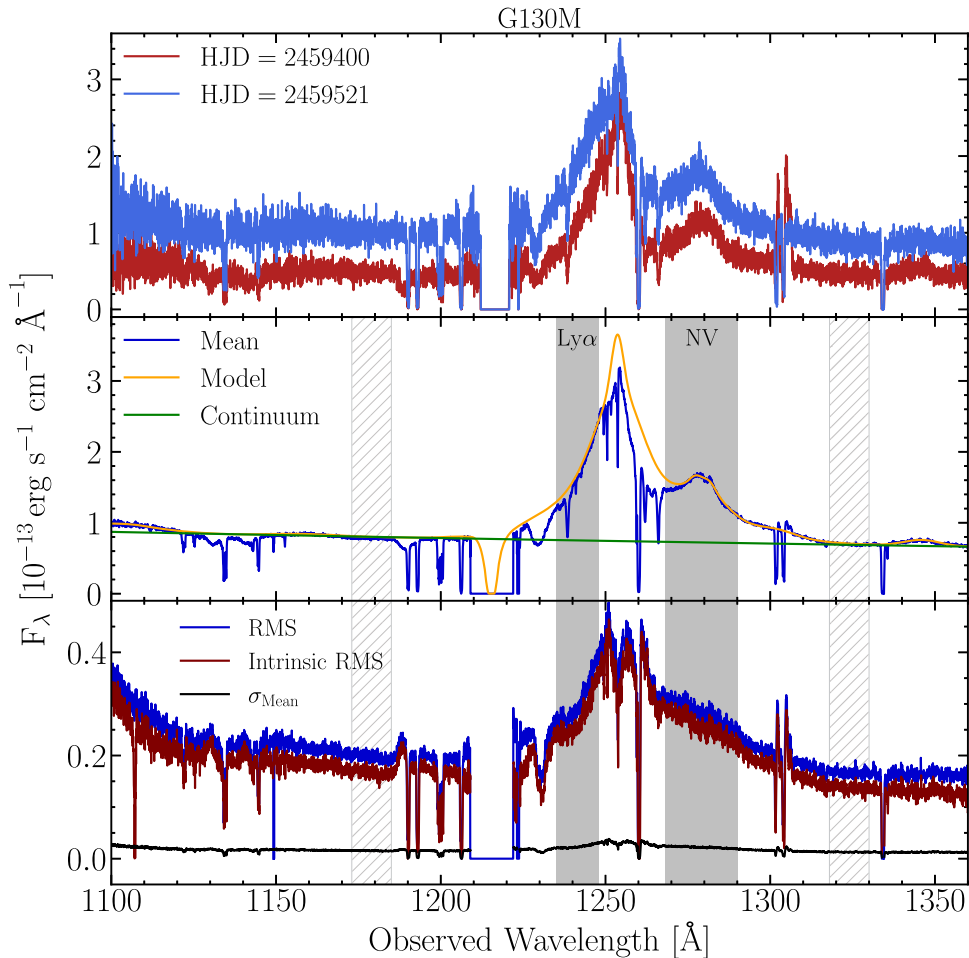


Figure 2. The top panel illustrates spectra from the epochs of lowest flux with strong absorption (red), and highest flux with weak absorption (blue). The red spectrum corresponds to HJD = 2459400, and the blue spectrum was obtained at HJD=2459521. The middle and bottom panels show the mean and rms spectra, respectively, for the 1100–1360 Å range of the G130M grating. The solid gray shaded regions show the integration windows for Ly α and NV, and the hashed gray regions illustrate the shortward and longward continuum windows used for continuum subtraction (see Table 2). The mean spectrum is computed from Equation (3). The empirical model of the intrinsic emission lines and continuum are superposed on the mean spectrum. In the bottom panel, the rms spectrum as defined in Equation (4) is shown in blue and the intrinsic rms spectrum is illustrated in red. The black line shows the total uncertainty in the mean, which is computed from the square root of the quadrature sum of statistical uncertainties (Equation (5)) and the uncertainties in precision of flux calibration (see Table 1) for the G130M grating.

absorption features associated with all permitted transitions in the spectrum. Kara et al. (2021) discuss the modeling approach adopted here in detail.

The heuristic spectral model for Mrk 817 is shown in the upper panels of Figures 2 and 3, superposed on the observed mean spectra. We chose continuum ranges that are least affected by absorption-line contamination and broad emission-line wings. Some of the emission lines (e.g., Ly α –NV and C IV–He II) have overlapping wings. In these cases, the boundary wavelength corresponds to the wavelength at which the fluxes of the two lines are comparable. We do not mask absorption lines at this stage in our analysis, so the regions chosen to measure the emission-line fluxes do not account for all flux in each of the lines. We report the total fraction that is included in our analysis for each line flux measurement in Table 2, where we integrate the flux in the mean model spectrum over the chosen wavelength regions. Detailed analysis of the overlapping emission lines requires in-depth modeling of individual emission and absorption lines as performed by Kriss et al. (2019), which will be provided in a separate paper.

We use the identified integration limits in Table 2 to measure the continuum fluxes using the weighted mean of the flux density in the integration region with weights defined as the inverse of the variance (square root of the propagated error).

Emission-line fluxes are measured as the numerical integral of the emission flux above a local continuum defined by a linear fit to the continuum flux in the continuum region adjacent to each emission line (see Table 2). The linear local continuum is then subtracted from the emission component. The line flux is numerically integrated over the integration limits given in Table 2 using Simpson’s method and excluding masked pixels. Statistical errors are computed numerically by creating $N_{\text{sample}} = 5000$ realizations of the line flux and the underlying linear continuum (see De Rosa et al. 2015). The flux, F_λ , in each realization is randomly generated from a Gaussian distribution having a mean equal to the flux of the spectral element and the width σ equal to the statistical error on the flux. For the linear continuum, we generate N_{sample} models using the best-fit values and covariance matrix of the linear fit to the data. For each realization, a line flux estimate is then obtained by subtracting the linear continuum and then performing the numerical integration of the residuals. The

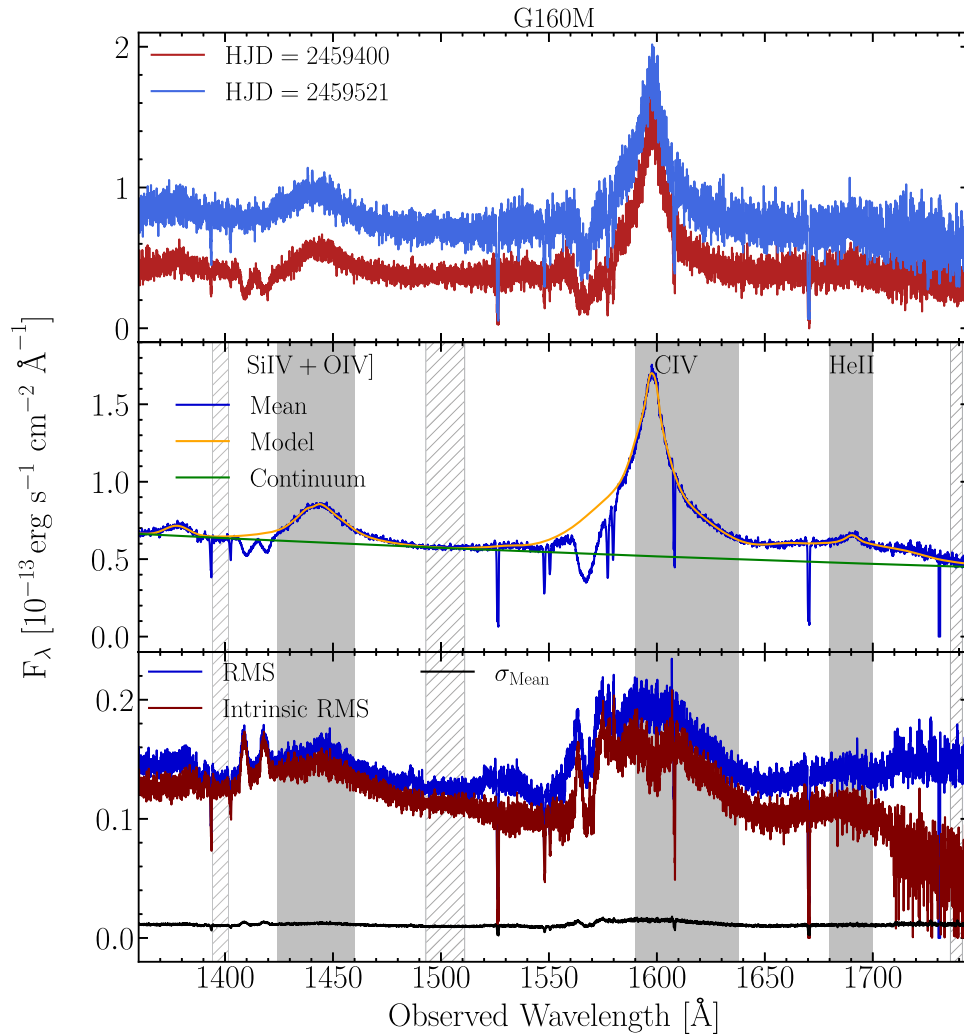


Figure 3. The top panel illustrates spectra from the epochs of lowest flux with strong absorption (red), and highest flux with weak absorption (blue). The red spectrum corresponds to HJD=2459400, and the blue spectrum was obtained at HJD = 2459521. The middle and bottom panels show the mean and rms spectra, respectively, for the 1360–1700 Å range of the G160M grating. The solid gray shaded regions show the integration windows for the Si IV + O IV blend, C IV, and He II emission lines, and the hashed gray regions illustrate the continuum windows used for each emission-line continuum subtraction (see Table 2). The mean spectrum is computed from Equation (3). The empirical model of the intrinsic emission lines and continuum are superposed on the mean spectrum. In the bottom panel, the rms spectrum as defined in Equation (4) is in blue. The intrinsic rms spectrum is in the red. The black line shows the total uncertainty on the mean, which is computed from the square root of the quadrature sum of statistical uncertainties (Equation (5)) and the uncertainty in precision of flux calibration (see Table 1) for the G160M grating.

1σ confidence levels are finally obtained from the distribution of the N_{sample} line fluxes. When the uncertainties are asymmetric, we adopt the larger of the two uncertainties as the statistical uncertainty associated with the integrated flux. Figure 4 shows the final full campaign continuum and line light curves. We report the light-curve statistics in Table 3. Tables 4 and 5 show a sample of the custom-calibrated light curves, which are available in machine-readable form in the online version of this paper.

The light curves in Figure 4 dramatically illustrate the high quality of the AGN STORM2 data set. The continuum variations are strong, with at least ten strong peaks more than 20% higher than the adjacent minima, and several with a factor ~ 2 contrast. Smaller variations on several-day timescales are readily visible, and they are well sampled by the two-day cadence and easily discernible given the small uncertainties ($<1.5\%$) in our processed data. These features make the differing characters of the continuum light curve and the emission-line light curves obvious. Early in the campaign, the continuum is near its brightest level, and it shows strong

intensity variations. In contrast, the emission-line light curves do not rise to a peak until a third of the way through the campaign, starting off at rather low levels, and gradually rising in flux. Also, despite the significant peak in the first ~ 50 days of the continuum light curve, the emission-line light curves are all relatively weaker. However, later in the campaign there are significant variations in line flux that do correlate with the continuum—the early emission-line light curves are not simply a smoothed, scaled, and shifted version of the continuum light curve. After the first third of the campaign, however, the emission lines track and correlate more closely with the continuum variations. Finally, during the last days of the campaign (2022 February 5–2022 February 24; HJD–2450000 9616–9634), the emission-line variations seem to decouple entirely from the continuum variations. This effect seems to be more prominent in the C IV emission-line light curve. This is possibly due to the appearance of a “BLR holiday” (Goad et al. 2016) at the end of the campaign. Even though the end of HST UV observations limited investigation of this event, the ground-based spectroscopic campaign is still ongoing and will

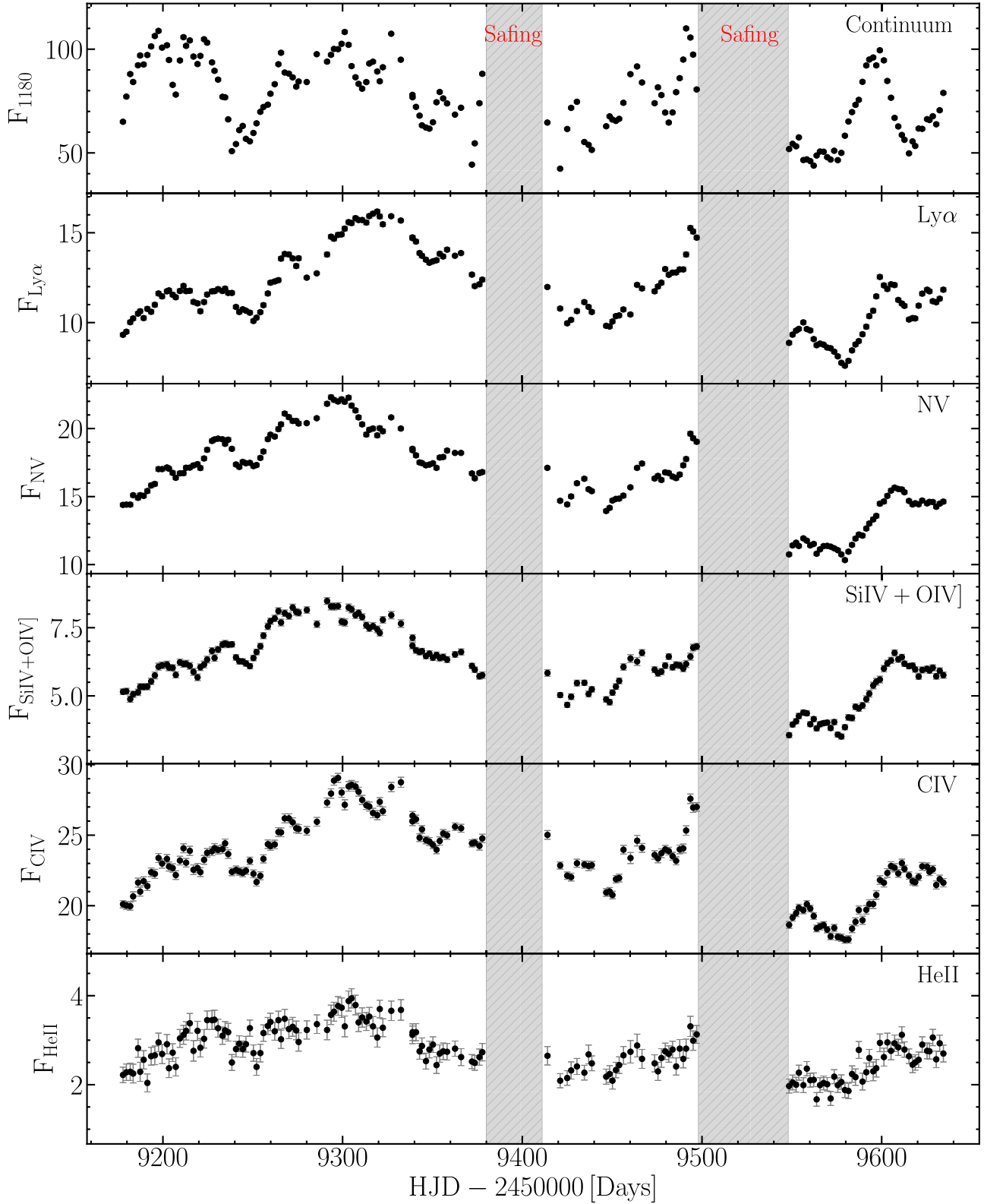


Figure 4. Integrated continuum light curve at 1180 Å (top panel) and emission-line light curves for Ly α , NV, Si IV + IV blend, CIV, and He II, respectively, in panels below the top panel. The Ly α light curve is the integrated flux in the blue wing of the emission line over the region not contaminated by absorption, and thus only includes 21% of the total emission-line profile (see Table 2). All the light curves are computed from the observed-frame spectra. The continuum light curve is in units of 10^{-15} erg s⁻¹ cm⁻² Å⁻¹ and the line fluxes are in units of 10^{-13} erg s⁻¹ cm⁻². The hashed gray regions display the two major observation gaps during our campaign, which were the result of HST safing events.

search for evidence of a BLR holiday at optical wavelengths. A thorough investigation of the light-curve behavior and individual reverberation windows will be discussed in an

accompanying paper (Y. Homayouni et al. 2023, in preparation). The reverberation lags between UV continuum bands will be described in a future paper in this series based on the

Table 3
Light-curve Statistics

Component	Mean and rms flux	Mean Fractional Error	F_{var}^a	Maximum Flux	Minimum Flux	Ratio ^b
$F_{\lambda}(1180)$	76.73 ± 17.69	0.01	0.23	109.99	42.4	2.59 ± 0.03
$F_{\lambda}(1398)$	62.62 ± 12.74	0.01	0.2	88.74	38.88	2.28 ± 0.02
$F_{\lambda}(1502)$	57.22 ± 11.63	0.01	0.2	79.84	34.36	2.32 ± 0.02
$F_{\lambda}(1739)$	46.35 ± 9.35	0.03	0.2	68.25	28.01	2.44 ± 0.11
Ly α	11.83 ± 2.04	0.01	0.17	16.18	7.6	2.13 ± 0.03
N v	16.53 ± 2.97	0.01	0.18	22.3	10.33	2.16 ± 0.03
Si IV + JO IV	6.11 ± 1.19	0.02	0.19	8.48	3.51	2.42 ± 0.07
C IV	23.21 ± 2.7	0.01	0.12	29.06	17.59	1.65 ± 0.03
He II	2.77 ± 0.49	0.07	0.16	3.94	1.67	2.36 ± 0.25

Notes. All light-curve statistics are reported in the observed frame. The continuum light curve at F_{1180} is in units of $10^{-15} \text{ erg s}^{-1} \text{ cm}^{-2} \text{ \AA}^{-1}$ and the line fluxes are in units of $10^{-13} \text{ erg s}^{-1} \text{ cm}^{-2}$.

^a F_{var} is defined as $(\sigma^2 - \delta^2)^{1/2} / \langle F \rangle$, where $\langle F \rangle$ is the mean of the observed flux, σ is the rms of the observed flux (second column), and δ is the mean statistical uncertainty (mean fractional error $\times \langle F \rangle$).

^b Ratio is the maximum divided by the minimum of the observed flux.

Table 4
Continuum Light Curves

HJD–2450000	F_{λ} (1180 $\text{\AA}$)	F_{λ} (1398 $\text{\AA}$)	F_{λ} (1502 $\text{\AA}$)	F_{λ} (1739 $\text{\AA}$)
9177.7825	65.01 ± 0.56	52.26 ± 0.34	48.07 ± 0.27	39.15 ± 1.20
9179.6703	77.14 ± 0.61	60.50 ± 0.37	54.52 ± 0.28	44.66 ± 1.27
9181.8164	88.00 ± 0.65	68.33 ± 0.39	60.57 ± 0.30	50.26 ± 1.33
9183.4419	84.18 ± 0.64	67.42 ± 0.39	59.37 ± 0.29	50.57 ± 1.33
9186.1714	92.24 ± 0.66	72.19 ± 0.40	63.84 ± 0.30	50.31 ± 1.33
9187.3528	96.96 ± 0.68	75.47 ± 0.41	66.23 ± 0.31	55.85 ± 1.38
9189.2933	92.61 ± 0.66	72.02 ± 0.40	65.39 ± 0.31	51.66 ± 1.34
9191.3455	97.19 ± 0.68	75.12 ± 0.41	66.27 ± 0.31	55.56 ± 1.38
9193.4633	101.32 ± 0.69	76.88 ± 0.41	69.08 ± 0.32	54.77 ± 1.38
9195.5147	106.34 ± 0.71	81.24 ± 0.42	71.99 ± 0.32	58.40 ± 1.41

Note. The continuum light curves are in units of $10^{-15} \text{ erg s}^{-1} \text{ cm}^{-2} \text{ \AA}^{-1}$. The full machine-readable table is available in the online journal article.

(This table is available in its entirety in machine-readable form.)

Table 5
Emission-line Light Curves

HJD–2450000	F (Ly α)	F (N v)	F (Si IV + JO IV)	F (C IV)	F (He II)
9177.7825	9.32 ± 0.10	14.39 ± 0.13	5.15 ± 0.10	20.11 ± 0.26	2.22 ± 0.17
9179.6703	9.49 ± 0.11	14.41 ± 0.14	5.17 ± 0.11	20.02 ± 0.28	2.28 ± 0.18
9181.8164	10.02 ± 0.11	14.41 ± 0.15	4.88 ± 0.12	19.97 ± 0.29	2.29 ± 0.20
9183.4419	10.23 ± 0.11	15.10 ± 0.14	5.06 ± 0.12	20.66 ± 0.32	2.25 ± 0.21
9186.1714	10.50 ± 0.11	14.91 ± 0.15	5.13 ± 0.12	21.64 ± 0.31	2.82 ± 0.20
9187.3528	10.63 ± 0.11	15.10 ± 0.16	5.33 ± 0.12	21.00 ± 0.32	2.29 ± 0.21
9189.2933	10.25 ± 0.12	15.05 ± 0.15	5.33 ± 0.12	21.74 ± 0.31	2.56 ± 0.21
9191.3455	10.76 ± 0.12	15.41 ± 0.15	5.34 ± 0.12	21.39 ± 0.32	2.04 ± 0.20
9193.4633	10.60 ± 0.12	15.82 ± 0.16	5.53 ± 0.12	22.37 ± 0.31	2.64 ± 0.20
9195.5147	10.99 ± 0.12	15.93 ± 0.15	5.75 ± 0.13	22.26 ± 0.32	2.66 ± 0.21

Note. The line fluxes are in units of $10^{-13} \text{ erg s}^{-1} \text{ cm}^{-2}$. The full machine-readable table is available in the online journal article. The flux measurements reported here are only a fraction of the total line flux due to contamination by absorption and blending (see the reported fractions in Table 2).

(This table is available in its entirety in machine-readable form.)

combined HST and Swift data. In the next section we analyze these light-curve properties more quantitatively. Note that the analysis below is focused on the HJD–2450000 9177–9615 time range and excluding the last nine epochs that might be affected by a BLR holiday.

4.3. Unusual Light-Curve Variability

Figure 5 shows the superposition of the 1180 $\text{\AA}$ continuum and C IV light curves. The continuum is initially strongly

variable while the emission-line light curves vary weakly and correlate poorly with the continuum; the continuum at the beginning of the campaign was near its peak levels, but the emission lines were not. This weak response and lower amplitude of variations is reminiscent of the “BLR Holiday” period of the STORM 1 campaign (Goad et al. 2016; Dehghanian et al. 2019a). However, after HJD 2459232 the emission-line light curves became well-correlated with the continuum. Similar periods of weak correlation reverting to a

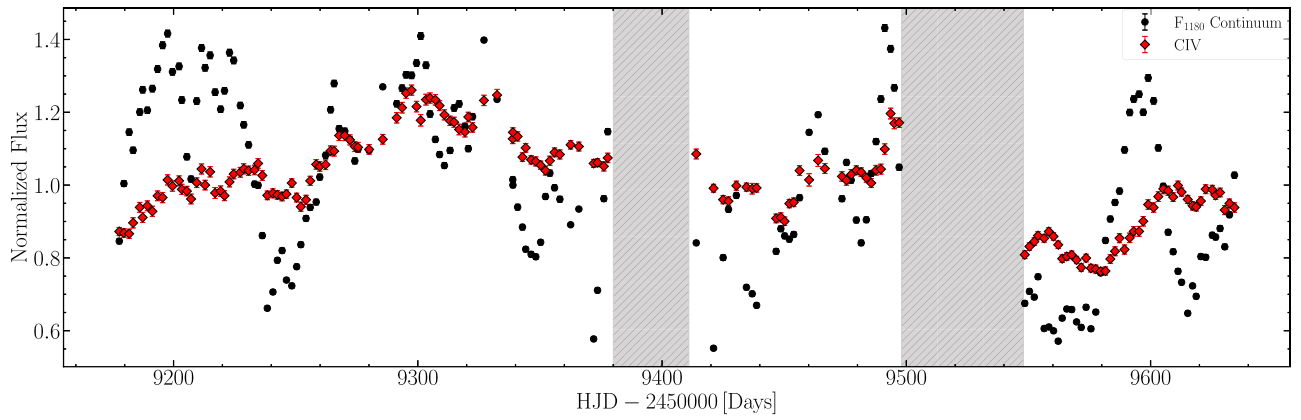


Figure 5. The 1180 Å continuum light curve (black) superposed on the C IV light curve (red). Each light curve is normalized to its median. Treating the year-long duration of the campaign as a single light curve reveals a time delay of $11.8^{+3.0}_{-2.8}$ days. However, the superposition of the continuum and line light curves show that a single time delay is an inadequate description of the light-curve behavior. In particular, note the dramatic continuum variability near HJD 2459230 and HJD 2459590 days, where the strong features in the light curves indicate a lag > 10 days, while away from these time periods, the light curves are more consistent with lags < 10 days.

strong response occurred throughout the remaining year of the campaign. These changes in the emission-line response to continuum variations leads to complexities in the lag measurement. In fact, the lag analysis presented in Kara et al. (2021) shows lags that are shorter than the measurements reported in this work (see Section 4). However, as we will discuss in upcoming work, the measured lag is a strong function of time interval analyzed. Further analysis of the light-curve variability and implications for the time delay is beyond the scope of this work, and will be presented in a subsequent paper.

5. Time-series Analysis

We use several widely-used time-series analysis methods for measuring reverberation lags: PyCCF (Sun et al. 2018), ZDCF (Alexander 2013), and JAVELIN (Zu et al. 2011). Below, we provide a brief overview of these lag measurement methods.

5.1. ICCF

The Interpolated Cross-correlation Function (ICCF) has been the most common method applied to previous RM studies (Gaskell & Sparke 1986; Gaskell & Peterson 1987; Peterson et al. 2004). ICCF calculates the Pearson coefficient r between two light curves shifted by a range of time lags (τ), using linear interpolation to match the shifted light curves in time, measuring r over the range of allowed τ . The centroid of the ICCF is computed using points surrounding the maximum correlation coefficient $r_{\max}$ ($r > 0.8 r_{\max}$), and the lag uncertainty is computed from Monte Carlo (MC) iterations of flux resampling and random subset sampling (Peterson et al. 1998). We implement ICCF using the publicly available PyCCF code (Sun et al. 2018) with an interpolation step of 1 day and a lag search range of ± 50 days. To estimate the uncertainties, we use 20,000 MC iterations of flux resampling and random subset sampling for PyCCF to generate a cross-correlation centroid distribution (CCCD), which gives the distribution of measured lags in all of the MC realizations. The upper and lower lag uncertainties are measured from the 16th and 84th percentiles of the primary peak. We report the lag centroid from the PyCCF method in Table 6. We also tabulate the FWHM for each emission line obtained from our fit to the mean spectrum. The FWHM is measured by taking the peak relative to the

Table 6
Emission Line Lags and Widths

Emission Line	Lag Centroid Days	Lag Peak Days	$r_{\max}$	FWHM (km s ⁻¹)
Ly α	$10.4^{+1.6}_{-1.4}$	$6.8^{+2.9}_{-1.9}$	0.61	3610
N V	$15.5^{+1.0}_{-4.8}$	$6.8^{+1.9}_{-1.9}$	0.64	3490
Si IV + JO IV	$8.2^{+1.4}_{-1.4}$	$6.8^{+1.9}_{-0.0}$	0.62	5220
C IV	$11.8^{+3.0}_{-2.8}$	$6.8^{+2.9}_{-1.9}$	0.61	3750
He II	$9.0^{+4.5}_{-1.9}$	$2.9^{+8.7}_{-0.0}$	0.67	7830

Note. The reported rest-frame time delays are based on the full campaign light curve (HJD 2450000 = 9177–9615) with a single lag measurement, and thus the complex response of the emission-line responses to the continuum light curve might not be sufficiently captured by it. The last column, FWHM, gives the full-width at half-maximum of each emission line in the mean spectrum.

continuum-subtracted line profile and measure the FWHM without excluding the negligible narrow-line contribution.

In this work, we only consider the continuum and line light curves for 2020 November 24 to 2022 February 05 (HJD 2459177–2459615), and exclude the last nine epochs from our lag analysis. We report the full campaign rest-frame time delays and uncertainties in Table 6 for all UV emission lines as indicators of the characteristic sizes of the regions that produced the observed emission-line flux variations. Figure 6 shows cross-correlation distribution functions from the PyCCF analysis along with the MC distribution of lag measurements for each emission line. Note that although all the CCFs have strong peaks with $r > 0.5$, they are very broad, likely indicating that the emission-line regions are spread over a large range in radial distance. Furthermore, our lag analysis reveals that the lags of the CCF peaks are consistently shorter than the CCF centroid lags for all the UV emission lines (see Table 6). This is caused by the asymmetry in the CCF, which may be an indication of an asymmetric transfer function or emission from an extended region (Cackett et al. 2007, 2022). We emphasize that the reported time delays in Table 6 are based on a single treatment of the light curve in the reverberation mapping scheme. Given the width and complexity of the CCFs, such a simple interpretation may not be an accurate description of the emission-line response to the varying continuum over the year-long campaign. We will discuss the implications of this

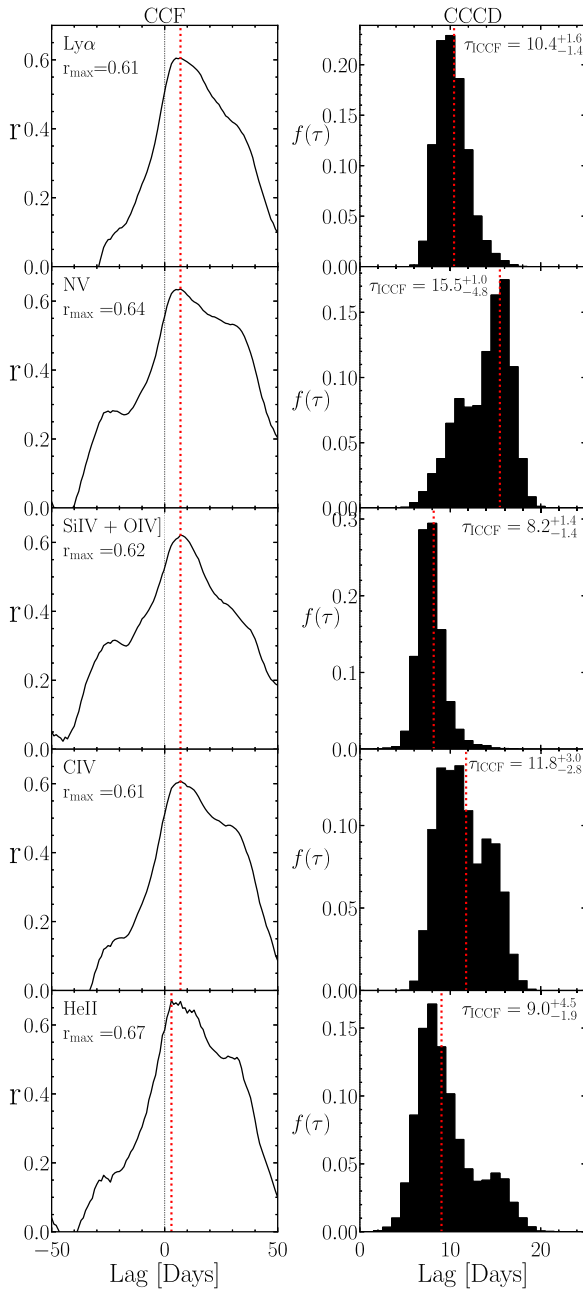


Figure 6. Results of the lag analysis using PyCCF for the emission-line light curves. For each F_{1180} and emission-line light-curve pair, we compute the cross-correlation coefficient with its maximum displayed by a red vertical line. The PyCCF CCCD are shown as black histograms, where $f(\tau) = N(\tau)/20,000$ is the total number of MC iterations normalized to unity. The maximum correlation coefficient and the lag measurement and uncertainties are displayed at the top-left corner of each panel.

behavior in an accompanying work (Y. Homayouni et al. 2023 in preparation)

5.2. Alternate Methods of Lag Measurement

In order to ensure that the time-delay measurements are not due to our choice of lag measurement method, we additionally used the ZDCF (Z-transformed discrete cross-correlation function; Alexander 2013) approach, in combination with a Gaussian process regression, to model the stochastic AGN light curves with arbitrary sampling (see Pancoast et al. 2015; Grier et al. 2017; Kovačević et al. 2018). For details on the approach,

see Kovačević & Popović (2017). We find that ZDCF yields similar time-delay measurements as PyCCF with time delays consistent within 1σ .

We also investigated whether a long-term trends in the light curve could affect these time-delay measurements (Welsh 1999). We used a running-median approach to remove the long-term trend and *detrended* the light curves. Using the detrended light curves, we measured the time delays. However, we found that a single trend covering the entire duration of our campaign does not fully capture the variability of the light curves. We will address more complex ways of detrending in a parallel work (Y. Homayouni et al. 2023, in preparation)

Finally, we also measure time delays using JAVELIN (Zu et al. 2011). JAVELIN uses a damped random walk (DRW) model to describe the stochastic variability of the AGN light curves. JAVELIN relies on the underlying assumption of physical reverberation: the emission-line light curves are scaled, time-delayed, and smoothed versions of the continuum light curve. JAVELIN uses a Markov chain Monte Carlo approach using a maximum-likelihood method to fit a DRW model to the continuum and emission-line light curves, assuming that the local accretion-disk response is a top-hat function and the reverberating light-curve model is the smoothed, scaled, and shifted version of the continuum light curve.

We allow the DRW amplitude to be a free parameter but fix the DRW damping timescale to 2000 days. Our campaign duration (~ 450 days) is much smaller than the typical damping timescale of an AGN (~ 1500 days in the observed frame; see Kelly et al. 2009; MacLeod et al. 2012). Thus, the damping timescale's exact value does not matter, so long as it is longer than the campaign's duration (the light curves are effectively modeled as a red-noise random walk with minimal damping). We also tested damping timescales of 500, 1000, 5000, and 10,000 days. Similar to the PyCCF method (Section 5.1), we measured the emission-line lags relative to the continuum F_{1180} variations. However, we found that a single DRW fit from JAVELIN to the entire duration of our campaign does not fully capture the variability of the light curve because as noted earlier, the line light curves are not simply smoothed and shifted versions of the continuum. The resulting model light curves do not track the data well in several time intervals of our campaign.

5.3. Velocity-binned Results

As discussed in Section 2.1, this program was designed to recover kinematic information about the BLR by resolving the emission-line response as a function of radial velocity. Similar to the AGN STORM 1 campaign on NGC 5548 (Horne et al. 2021), the full analysis of velocity-delay maps will be presented in a subsequent paper in the AGN STORM 2 series. Here we carry out a preliminary analysis intended to show whether velocity-dependent information is present in the data. We isolate the Ly α and CIV profiles as described earlier in Section 4.2 and integrate the fluxes in bins of width 250 km s^{-1} , probing $-7500 \text{ km s}^{-1} < \Delta V < 4000 \text{ km s}^{-1}$ for Ly α , and $-9000 \text{ km s}^{-1} < \Delta V < 11,000 \text{ km s}^{-1}$ for the CIV emission line.

Figures 7 and 8 show velocity-dependent lags for Ly α and CIV, respectively. The Ly α emission line shows a strong velocity-dependent profile. However, blending with NV on the red wing, and the presence of multiple broad absorption features, contaminate the Ly α emission-line flux and thus affect

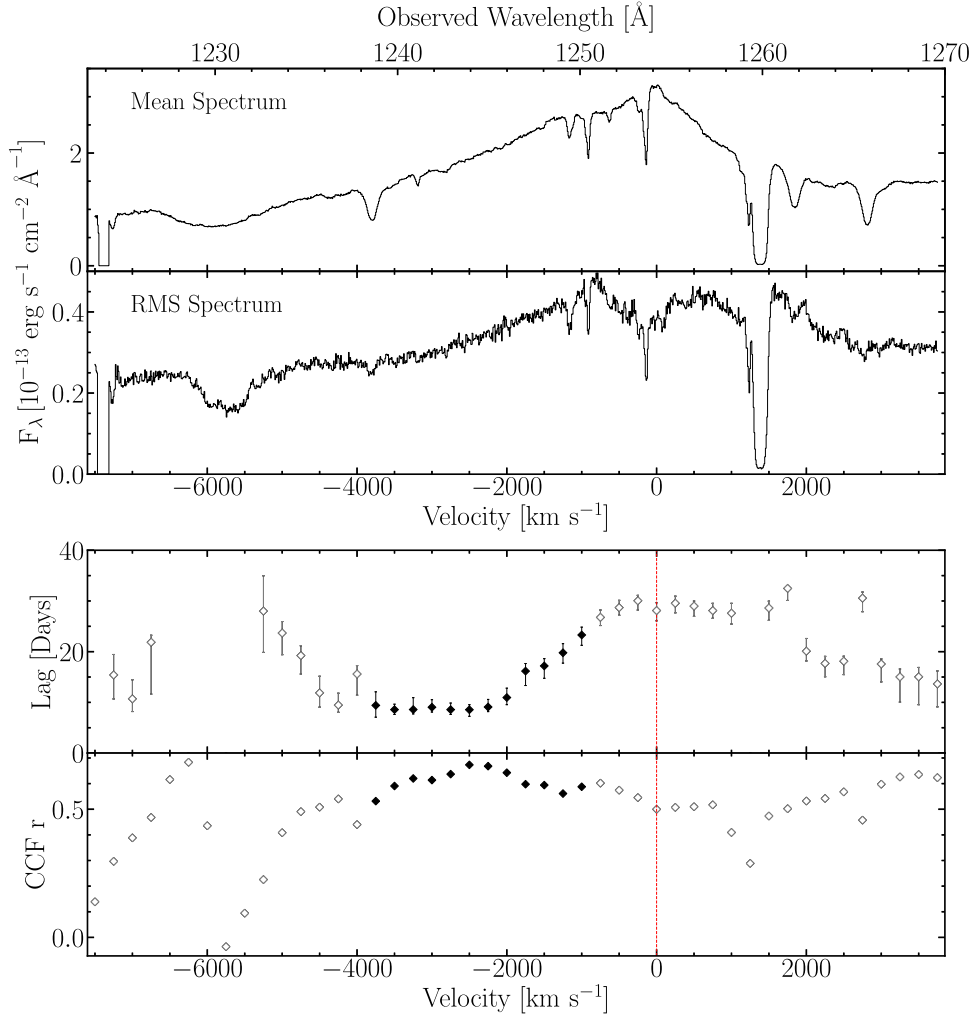


Figure 7. Rest-frame velocity-dependent time lags for the Ly α emission line. The top two panels show the Ly α mean and rms line profile respectively. The bottom two panels show the distribution of velocity-binned lags for each 250 km s^{-1} bin and the associated maximum cross-correlation value r . The Ly α emission-line profile is highly contaminated by several broad absorption features both at the line center (due to N V) and in the red wing (as indicated by open symbols). However, the blue wing shows a clear signature of virialized Ly α gas where high-velocity gas responds more rapidly, and low-velocity gas that is toward the line center responds later. Detailed absorption models are needed to extract the underlying Ly α emission-line profile and recover the velocity-dependent lag results, which will be the subject of future work in the series.

the results (see open symbols in Figure 7). Many of the computed lags that are in absorption-contaminated bins are associated with small correlation coefficients and/or negative lags and are not included in the figure. We find small lags for the outer blue wing of the velocity profile, indicating that high-velocity gas responds more rapidly than the low-velocity gas near the line center. This general pattern is similar to what has been seen in other objects and is consistent with a virialized region. However, this virialized motion is mostly present in the central region of the Ly α profile, and the complex response of emission lines to continuum variations as characterized in Figures 4 and 5 could indicate that other gas motions may also be present in the BLR, but are less dominant near the center of the line profile. Full modeling of the spectra to remove the impact of absorption and the blending of the emission lines (see, e.g., STORM VIII Kriss et al. 2019), and different approaches to producing 2D velocity-resolved reverberation maps (see e.g., STORM IX Horne et al. 2021, and STORM XII Williams et al. 2020) will elucidate the gas dynamics in the BLR.

The C IV emission line shows strong velocity-dependent structure that spans a wide range of time lags. We have similar

problems due to absorption in the blue wing ($<1580 \text{ Å}$) of the C IV profile. The central portion of C IV shows a distinctive “M”-shaped profile, characteristic of a thick rotating disk (Welsh & Horne 1991). The red wing, which is free of absorption, shows a virial profile with long lags near the line center and shorter lags at higher velocities. To investigate the robustness of the velocity-binned lag results, we also show the maximum correlation coefficient profiles in the Ly α and C IV emission-line profiles in the bottom panels of Figures 7 and 8. Velocity bins showing poor correlation ($r < 0.4$) are more frequent in windows that are affected by absorption. We need detailed modeling of the broad absorption features to recover the uncontaminated emission-line profiles and refine our velocity-binned analysis. This will be the subject of subsequent work in the AGN STORM 2 series of papers.

6. Discussion

Mrk 817 has been observed in previous ground-based optical campaigns (Peterson et al. 1998; Denney et al. 2010) with multiple results for the H β mean lag and velocity-binned

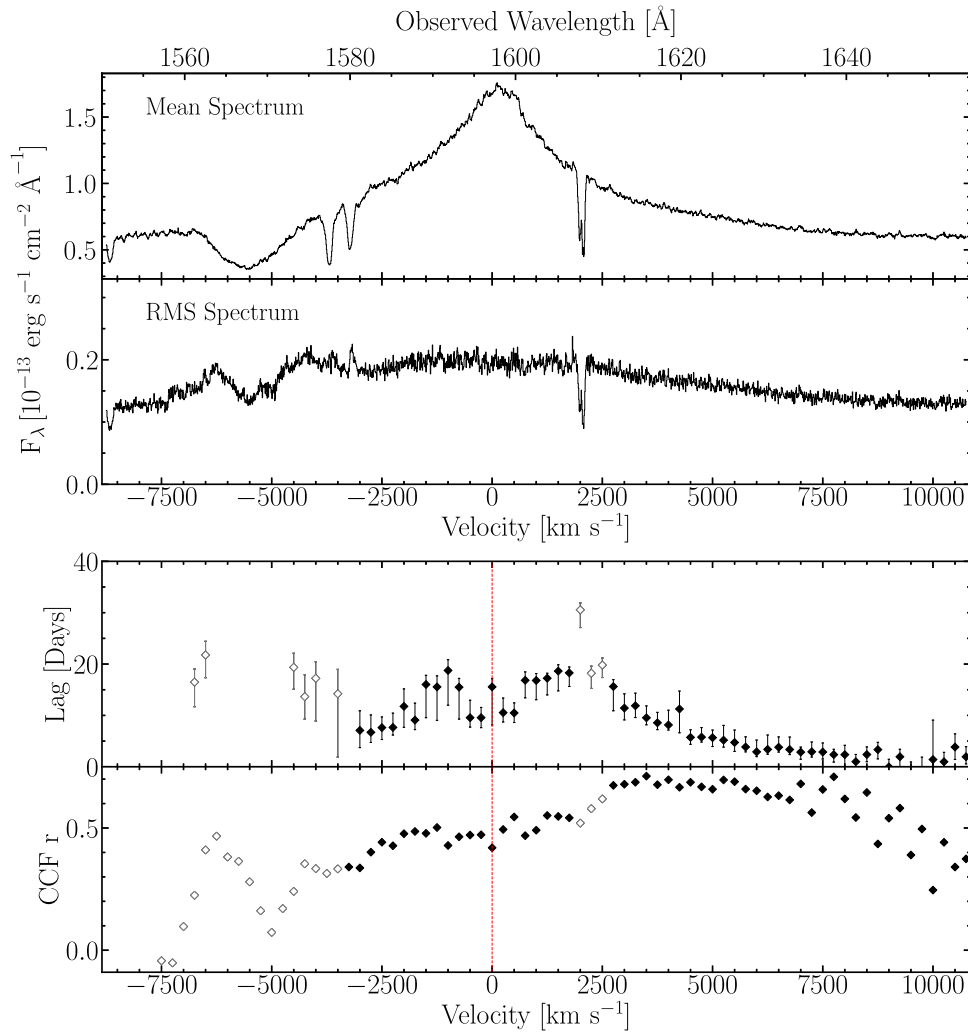


Figure 8. Rest-frame velocity-dependent time lags for the C IV emission line. The top two panels show the C IV mean and rms line profile, respectively. The bottom two panels show the velocity-binned lags for each bin and the associated of the maximum cross-correlation coefficient r . The open symbols illustrate where the spectra has been contaminated by absorption. The lag velocity profile has an “M” shape, with a local minimum near the line center, local peaks at $\sim \pm 1500 \text{ km s}^{-1}$, and the shortest lags at higher velocity in the wings.

analysis. However, the current campaign provides the only velocity-binned results on Mrk 817 in the UV. Even though the target was primarily selected because of absence of absorption in its emission-line core and wings, the first HST observation revealed the presence of strong broad and narrow absorption lines, which complicates measurement of the true emission-line fluxes. Nevertheless, we find significant variability in the continuum and emission-line light curves. We are able to measure BLR lags for prominent high-ionization lines that have not been observed before in Mrk 817.

Our analysis of the velocity-binned lags reveals that the C IV line shows a clear velocity-resolved response. Although broad absorption contaminates the far-blue wing of the C IV profile, it does not overlap with the central portion of the emission line, where we recover a strong velocity-dependent signal. The C IV line has an “M”-shaped velocity profile, with a local minimum (~ 10 days) near the line center and peak at longer lags (~ 20 days) at roughly $\pm 1500 \text{ km s}^{-1}$, which is indicative of a thick rotating disk (Welsh & Horne 1991). Similar profiles have been found in NGC 5548 for Ly α and C IV (Kriss et al. 2019), and also for H β (De Rosa et al. 2018).

The Ly α profile is highly contaminated by absorption troughs or blended with NV. However, in the narrow, contamination-free window of the Ly α blue wing, we find kinematic signatures that are consistent with an extended BLR in virial motion.

Although we can measure time lags for all the prominent UV emission lines, the observed light curves are not simply the smoothed, scaled, and shifted versions of the continuum light curve. This complicated response of the BLR to the continuum variations is likely associated with the broad and narrow absorption features in the UV spectra, and to the heavy obscuration in the X-rays (Kara et al. 2021). This absorbing material can shield the BLR from the ionizing continuum, making the observed UV continuum an imperfect proxy for the true ionizing flux.

The AGN STORM 2 data includes UV emission lines with a wide range of ionization potentials and probe a broad range of lag measurements (Table 6). Typically, we expect shorter lags for emission lines with higher ionization potentials. Examining the centroids of the lag distributions in Table 6, however, we see that this sequence is completely reversed. Lines with the highest ionization potentials (NV and He II) have the longest

lags, and lower-ionization ions such as Si IV + JO IV have the shortest lags. This unexpected relationship is another reflection of the complexity of the BLR structure we have uncovered in the campaign. However, the interplay between ionization state and thermalization density could introduce deviations from this expected relation. For example, despite its relative small formation ionization potential, Si IV 1398 Å has a surprisingly small expected emissivity-weighted radius, comparable to that of JO IV 1402 Å (see Korista & Goad 2019 for more detail). Similarly, Hu et al. (2020) found complicated structure in PG 2130 +099 between the ionization potential and the lag for He I, H β , and Fe II.

In conclusion, the BLR gas distribution is more complicated than can be captured using a single lag measurement (see Figure 6). The CCFs for all the lines in Figure 6 are noticeably asymmetric, and the peaks of the distributions, also shown in Figure 6, generally show shorter lags than the centroids. Using this measure, the shortest lag is for the high-ionization line He II, with all other lines having only a slightly longer peak lag. Given these complexities, we are not able to analyze fully the BLR ionization stratification in this work, and we leave a more detailed examination of the complex behavior of the BLR response to a parallel paper (Y. Homayouni et al. 2023, in preparation).

Preliminary results from this analysis show the potential to reveal the types of structured maps that will provide additional constraints on future models of the BLR, and more clearly reveal distinct kinematic structures responsible for the velocity-resolved signatures we presented here.

7. Summary

We present the first results from 165 epochs of HST COS observations as part of the multiwavelength AGN STORM 2 campaign on Mrk 817 to measure mean and velocity-dependent RM time lags for broad UV emission lines. Mrk 817 was observed with an average 2 day cadence from 2020 November 24 to 2022 February 24 (HJD 245000 = 9177–9634). This work reports the final custom calibration of the COS data with homogeneous reduction that have a local flux precision <1.5%. Our major findings are as follows.

1. The UV continuum at 1180 Å and the emission-line light curves for Ly α , N V, Si IV + JO IV, C IV, and He II show significant variability.
2. HST observations from the last 9 epochs point to a plausible BLR holiday. Ongoing ground-based observations may be able to reveal whether the BLR holiday also extends to the optical regime.
3. We report the mean time-delay measurements for all UV emission lines (see Table 6) with respect to the 1180 Å continuum. However, we find that a single lag measurement is an inadequate description of the light curves. The emission-line responses are not simply a smoothed, scaled, and shifted version of the continuum variations. We will discuss the complex behavior of the BLR response in an accompanying paper (Y. Homayouni et al. 2023, in preparation).
4. Our qualitative analysis of the velocity-dependent lags only focuses on Ly α and C IV owing to significant contamination of the rest of the emission-line profiles. Velocity-dependent lags in the blue wing of Ly α are consistent with virial motion, with the shortest lags

present in the high-velocity wings, and the longest lags in the lower-velocity center. The velocity-resolved analysis for C IV instead reveals an “M”-shaped profile hinting at BLR gas in a thick rotating disk.

5. There is no simple connection between emission-line ionization potentials and their BLR lags. The BLR gas and the emission-line response are more complex than can be reflected by a single lag measurement.

We plan to use our custom-calibrated HST light curves for the analyses in upcoming work. Even though the HST portion of the campaign has ended, monitoring of Mrk 817 is continuing with Swift, NICER, and ground-based observations.

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