Solving the puzzle of discrepant variability on monthly time scales implied by SDSS and CRTS datasets

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ABSTRACT

We present improved error analysis for the 3,800 CRTS (Catalina Real-Time Transient Survey) optical light curves for quasars from the Sloan Digital Sky Survey Stripe 82 catalog. SDSS imaging survey has provided a time-resolved photometric dataset which greatly improved our understanding of the quasar optical continuum variability: data for monthly and longer timescales are consistent with a damped random walk. Recently, newer data obtained by CRTS provided puzzling evidence for enhanced variability, compared to SDSS results, on monthly time scales. Quantitatively, SDSS results predict about 0.06 mag root-mean-square variability for timescales below 50 days, while CRTS data show about a factor of two larger rms for spectroscopically confirmed SDSS quasars. Our analysis presented here has successfully resolved this discrepancy as due to slightly underestimated photometric uncertainties provided by the CRTS image processing pipelines. The photometric error correction factors, derived from detailed analysis of non-variable SDSS standard stars that were re-observed by CRTS, are about $20-30\%$, and result in a quasar variability behavior implied by the CRTS data that is fully consistent with earlier SDSS results.

1 INTRODUCTION

Variability can be used to both characterize and select quasars in sky surveys . Athough various timescales of variability can be linked to physical parameters, such as accretion disk viscosity, or corona geometry (Kelly+2011, Graham+2014), the physical mechanism remains elusive, and most viable explanations include accretion disk instabilities (Kawaguchi + 1998), surface thermal fluctuations from magnetic field turbulence (Kelly+2009), coronal x-ray heating (Kelly+2011) (see Kozlowski 2016 for review). The diversity of physical scenarios explaining the origin of quasar variability led to a plethora of ways to characterize the varying brightness. The two most widely used approaches to describe variability of quasars are damped random walk (DRW) and structure function. The DRW model is more suited for lightcurves with typical cadence of days (Zu+2013, Kozłowski+2016), whereas an ensemble SF analysis is better for sparsely sampled lightcurves (Hawkins 2002, Vanden Berk 2004 , de Vries 2005, Schmidt 2010, Graham 2014, or review in Kozlowski 2016). Although CRTS data has been used for both an SF and DRW analysis [CITE WHERE], we use the SF approach, as more robust for sparsely sampled lightcurves than the DRW. [For a recent overview of the context for variability studies, see Lawrence+2016.]

Although SF can be defined in a variety of ways, it can be characterized by a simple broken power law (Schmidt+2010). At short timescales, the variability amplitude increases, following a steeper power law index, until the power law index starts to flatten above the characteristic timescale τ. This knee in the power law description may correspond to a transition from the stochastic thermal process driving the variability, to the physical response of the disk that successfully dampens the amplitude on longer timescales (Lawrence+2016, Kelly+2007,2009,2011, Collier+Peterson 2001 (linking the amplitude to the black hole mass)). Indeed, in x-rays quasars are described by a broken power law, where the break timescale is linked to the size of x-ray emitting region (Kelly+2011 in Graham+2014).

Although previous studies found that $\tau > 100$ days (MacLeod+2011, Kozłowski+2016), recently, Graham+2014 used the SF, DRW, and Slepian Wavelet Variance (SWV) analysis for CRTS and SDSS S82 lightcurves. Using SWV on CRTS data they found characteristic time scales at QSO rest frame of $\log_2(\tau) = 5.75$) 54 days, and $(\log_2(\tau) = 8.2)$ - 294 days. Additionally, using this method on S82 data they found a peak at $\log_2(\tau) = 7.25$, and for OGLE at $\log_2(\tau) > 6$. The short timescale of $\tau = 54$ days is surprising, as it is shorter by a factor of two than any previous estimates (MacLeod+2011, Zu+2013). We set out to reana-

lyze the CRTS data, and investigate the plausibility of this discrepant timescale.

2 DATA SETS

We study stars and quasars from Stripe 82, using the Sloan Digital Sky Survey and the Catalina Real-time Transient Survey data. Stripe 82 is a large (over 100 $\deg^2$), repeatedly observed reqion of the equatorial sky ($22^h24^m < \mathrm{RA} < 04^h08^m$ and $|\mathrm{Dec}| < 1.27 deg$) (Sesar+2007, Suveges+2012).

2.1 Sloan Digital Sky Survey (SDSS)

We use the SDSS catalog data with robust, five-band near-simultaneous photometry for 9258 quasars, and 1006849 standard stars - we use it for photometric color information and sample selection. The quasar catalog contains spectroscopically confirmed quasars from the SDSS Data Release 7 (Abazajian+2009), based on SDSS Quasar Catalog IV (Schneider+2008, VizieR Online Data Catalog, 7252, 0), and was compiled by Macleod+2011 (see [1]) . The standard stars catalog ver. 2.6 was compiled by Ivezic+2007 (see [2]) .

2.2 Catalina Real-time Transient Survey (CRTS)

The CRTS data consists of white light (no filter) lightcurves - it was designed to find near-Earth objects, hence short intra-night cadence, to allow a rapid follow up (Graham+2015). Three survey telescopes (0.7m Catalina Sky Survey Schmidt in Arizona, 1.5m Mount Lemmon Survey telescope in Arizona, and the 0.5m Siding Spring Survey Schmidt in Australia) were equipped with identical, 4kx4k CCDs (see Djorgovski+2011 for technical details CRTS), taking 4 exposures each night. Although in principle white light magnitudes can be calibrated to Johnson V-band zero point (Drake+2013), we found it a redundant step for our analysis.

For our study we used a sample of 7932 spectroscopically confirmed S82 quasars, prepared by B. Sesar, from the Data Release 2, based on the list by MacLeod+2011. The majority (96%) of CRTS quasar lightcurves span the time of 7-9 years, with a mean sampling of 1 to 4 times per night, on 70 nights on average (see Fig.1, upper-left panel). Mean interval between epochs is 209 days (Fig.1 bottom-right panel), and the mean quasar magnitude is 19.5 mag.

Our comparison sample consists of CRTS lightcurves of 52133 randomly chosen 10% of the S82 standard stars catalog ver.2.6 (Ivezic+2007), extracted by B. Sesar from the CRTS DR2.

2.3 Catalog Matching

To enrich the information about each CRTS object with SDSS color information, we matched positionally CRTS data to SDSS data within S82 using astropy

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`match_to_catalog_sky` routine[3]. Given that the SDSS catalog has 100 times more stars than CRTS, we found an SDSS counterpart to all CRTS stars within 0.01 arcsec matching radius. However, since there is less SDSS quasars in the S82 catalog, and we found an SDSS counterpart to 7586 CRTS quasars within 0.36 arcsec. We ignored 15 quasars for which no SDSS counterpart was found within 1 arcsec.

2.4 Preprocessing

Because we are not investigating hourly timescales, prior to structure function analysis we day-averaged all CRTS light curves. Similarly to Charisi+2016, we replaced the magnitudes from the *j*-th night by their mean weighted by the inverse square of error:

$$m_j = \langle m_{ij} \rangle = \frac{\sum_{i=1}^{N} w_{i,j} m_{i,j}}{\sum_{i=1}^{N} w_{i,j}} \qquad (1)$$

where $i = 1...N$, is the number of observations per night, with weights $w_{i,j} = e_{i,j}^{-2}$. We estimate the error on the weighted mean by the inverse square of the sum of weights:

$$\sigma_j(m_j) = \frac{1}{\sqrt{\sum_i w_{i,j}}} \qquad (2)$$

[any CITATION ? is it a standard thing to do ?] Finally, to avoid unphysically small errors, we added in quadrature 0.01 mag to σ_j if $\sigma_j < 0.02 mag$.

2.5 Selection

We selected both quasars and stars using a combination of information from SDSS and CRTS. To find magnitude difference between different days we first required that the raw lightcurve has more than 10 epochs, which from initial 52131 stars and 7932 quasars left 49385 stars and 7707 quasars. We thus also removed those lightcurves with less than 10 days of observations, leaving 7601 quasars and 48250 stars. We also required that the lightcurve-average of day-error ($\langle \sigma_j(m_j) \rangle$) be less than 0.3^{mag}. Since the raw distribution of errors peaks at lower values (mean of 0.19 for stars and 0.22 for quasars) than the distribution of the weighted mean errors (mean of 0.13 for stars and 0.15 for quasars), this cut only removes less than 10% of lightcurves. Our final sample consists of 7108 quasars and 42864 stars.

3 ANALYSIS METHODS

To analyze the quasar and stellar lightcurves we consider a relationship between measured data m_j at times t_j, and its "copy" shifted by Δt (Kozłowski+2016). We bin the data in bins of Δt, and calculate statistics that characterize the variability of stars and quasars. By splitting our sample of stars by color into "blue" and "red" we compare their variability properties to those of quasars in three magnitude bins.

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[1] http://www.astro.washington.edu/users/ivezic/cmacleod/qso_dr7/Southern.html
[2] http://www.astro.washington.edu/users/ivezic/sdss/catalogs/stripe82.html

[3] http://docs.astropy.org/en/stable/coordinates/matchsep.html#matching-catalogs

3.1 Structure Function

The structure function is a well-studied approach to characterizing lightcurves (Vanden Berk +2004, de Vries+ 2005, Kozłowski+2016, Graham+2015) . To avoid the uncertain redshift estimate based on SDSS spectra that would be required to correct to the rest-frame variability, we use the observed frame time lags (like Schmidt+2010) (see Kozlowski+2016). The magnitude difference $\Delta m_{j,k}$ for $\Delta t_{j,k} = |t_j - t_k|$, and the errors added in quadrature : $\sigma_{j,k}^2 = \sigma_j^2 + \sigma_k^2$.

We prepared 'master' files for stars and quasars, with $\Delta m_{j,k}$, $\Delta t_{j,k}$, $\sigma_{j,k}$. After day-averaging, given that the median lightcurve length is 70 days, each object on average has $\sum_{i=2}^{70}(i-1) = 2415\ 2E3$ pairwise magnitude difference points. Therefore selecting to study eg. 1000 Quasars, we have approximately 2E6 magnitude differences. We study statistical properties of magnitude differences vs time lag by calculating various properties in linearly spaced bins of $\Delta t_{j,k}$. We calculated statistical descriptors for each bin, as described in (SECTION ABOUT STATISTICS). We found that binning does not affect significantly the shape of structure function nor our conclusions about error properties, and that choosing 200 bins was optimal for our purpose.

We characterized the distribution of $\Delta m_{j,k}$ vs $\Delta t_{j,k}$ by calculating statistics for each bin: the rms standard deviation $\sigma_{stdev} = \sqrt{(mean(abs(x - x.mean()) ** 2))}$, the standard deviation based on the interquartile range: $\sigma_G = 0.714IQR$, where IQR = 25

For each delta(tau) bin, to calculate sigma and mu we analysed a distribution of approximate values of those parameters for 1000 boostrapped resamples of the (m_{ij}, e_{ij}) data. The maximum and minimum of this distribution served as upper and lower bound on calculation of the full solution. We find that, as in AstroML Fig. 5.8, approximate calculation slightly underestimates sigma, and thus we chose to use full solution, based on marginalizing the log-likelihood of various values of sigma that could describe the (m_{ij}, e_{ij}) ensemble. [DESCRIBE IN MORE DETAIL, AS IN PAPER II ?]

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