## LSST AGN SC Cadence Note: Two metrics on AGN variability observables

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### 1. EXECUTIVE SUMMARY

We have developed two metrics related to AGN variability observables (time-lags, periodicity, and Structure Function (SF)) to evaluate LSST OpSim FBS 1.5, 1.6, 1.7 performance in AGN time-domain analysis. For this purpose, we generate an ensemble of AGN light curves based on AGN empirical relations and LSST OpSim cadences. Although our metrics show that denser LSST cadences produce more reliable time-lag, periodicity, and SF measurements, the discrepancies in the performance between different LSST OpSim cadences are not drastic based on Kullback-Leibler divergence. This is complementary to Yu and Richards results on DCR and SF metrics (see Yu's talk https://docs.google.com/presentation/d/12Q1zKiWtoQAXsh7GS6J9TYhEKsWfkN5sVDufEyCKkEE/edit#slide=id.gc6954dd1ce\_0\_3), extending them to include the point of view of AGN variability.

## 2. THE TWO METRICS

A detailed description of metric analysis performed on OpSim 1.5 and 1.6 is given in Kovačević et al. (submitted version can be found in https://github.com/LSST-sersag/white\_paper/blob/main/data/paper.pdf). Results based on the OpSim 1.7 are available at https://github.com/LSST-sersag/agn\_cadences.

## 2.1. Metrics based on AGN time lag $(\tau)$ and periodicities (P) measurements

We designed a metric  $\log \phi_{\mathcal{T}}$  for time-lags and periodicities that identifies smooth trends (i.e. hyperplanes or linear relations) in complex datasets. This scaling relation ties relative formal error of measured time scale  $\mathcal{T}_{\text{obs}} \in (\tau, P)$  to the characteristics of AGN light curve and OpSim cadendes: the amplitude of flux variation  $F_{\text{var}}$  relative to the measured flux (or photometric) error  $\sigma$ :  $(\frac{F_{\text{var}}}{\sigma})$ ; the ratio of observed time scale  $\mathcal{T}_{\text{obs}} \in (\tau, P)$  and light curve sampling time  $\Delta t$ :  $\frac{\mathcal{T}_{\text{obs}}}{\Delta t}$ . The proposed metric is:

$$\log \phi_{\mathcal{T}} = \log \frac{\sigma_{\mathcal{T}}}{\mathcal{T}} \propto A + C_1 \frac{F_{\text{var}}}{\sigma} + C_2 \frac{\mathcal{T}_{\text{obs}}}{(1+z)\Delta t},\tag{1}$$

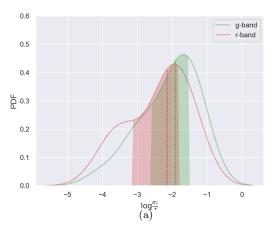
where we assume that the error of the rest-frame time lag or periodicity inferred from the observed light curve will be increasing with an increasing redshift of the object for the same observed time scale and sampling time. The metric coefficients tell us how to compare variables on a given ensemble of objects. Measuring the observables and their uncertainties for each object, we can estimate metric (i.e., coefficients  $A, C_1, C_2$ ) for given AGN light curves ensemble generated from OpSim cadences and probability density function (PDF) of the metric.

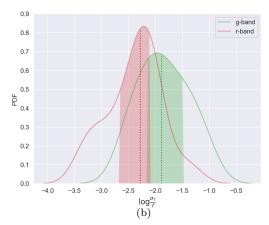
### 2.2. Structure function metric

The idea of a metric is to estimate deviations of SFs based on OpSim cadences (termed "gappy" light curve,  $SF_{\rm gappy}$  from 'reference' SF obtained from densely and uniformly sampled light curve ( $SF_{\rm conti}$ ). Let an ensemble of k simulated light curves be generated, providing  $SF_{\rm conti}^i$  and  $SF_{\rm gappy}^i$ , i=1,k where k is the number of simulated light curves. Then simplistic metric of similarity between  $SF_{\rm conti}^i$  and  $SF_{\rm gappy}^i$ , i=1,k curves is given by  $M^i=(SF_{\rm conti}^i-SF_{\rm gappy}^i)^2$ , where  $M^i$  is the ensemble of the deviations curves of SF. However, we can generalize this concept. We calculate these deviations curves for objects in different redshift bins and average the deviations curves for each redshift bin z:

$$M_z^2 = \frac{1}{N_z} \sum (SF_{\text{conti}}^i - SF_{\text{gappy}}^i)^2 = \frac{1}{N_z} \sum_{i=1}^{N_z} M_i^2,$$
 (2)

where  $M_z^2$  is an averaged deviation curve for redshift bin z,  $N_z$  is the number of deviations curves  $M^i$  within the redshift bin z. We can plot thesse averaged SF deviations curves on the redshift and characteristics time scale domain as 2D maps or to calculate PDFs on both domains.





**Figure 1.** Densities of metrics for the time lags (a) and periodicities (b) measured on the artificial set of objects (see an auxiliary Table 1 at <a href="https://github.com/LSST-sersag/white\_paper">https://github.com/LSST-sersag/white\_paper</a>). Vertical dashed lines denote the medians of distributions, and shaded regions correspond to ranges of 25 and 75 percentiles.

### 3. ANALYSIS OF RESULTS

We tested our metrics on simulated AGN light curves. More details in Kovačević et al. and on the GitHub (https://github.com/LSST-sersag/white\_paper/blob/main/data/paper.pdf)

# 3.1. Metrics based on time lags and periodicities

We learned the time lags and periodicities metrics and estimated probability density (PDF) of performances of these metrics for simulated objects based on g and r-band LSST OpSim cadences (see auxiliary Table 1 in https://github.com/LSST-sersag/white\_paper/blob/main/data/table\_1.pdf). How much given distributions are away from each other we estimated using Kullback-Leibler (KL) Divergence  $D_{KL}(pdf1||pdf2)$ .

Figure 1(a) displays the PDF of time lag metrics for g and r-band objects. Here the KL divergence is almost symmetric  $D_{KL}^{\tau}(pdf_g||pdf_r) \sim D_{KL}^{\tau}(pdf_r||pdf_g) \sim 0.02$  indicating that our metric detected a slight difference of time-lag measurement in g and r-band. A simple visual inspection of the plot indicates that the smaller relative formal errors (and their median) are more likely to occur for the r- band light curves. However, metrics for periodicities measurement (see Fig. 1(b)) are more sensitive to sampling density as KL- divergence is two times larger  $D_{KL}^{P}(pdf_g||pdf_r) = 0.04$ . The medians of metric performance in two bands are at a larger mutual distance than those for time lags. Similar to time lags, more reliable oscillation detection is expected in the r-band. This is not surprising since the r-band contains almost three times denser cadences, enabling light curves to capture more information from underlying oscillations. The DDF cadences are very likely to produce confident time-lag and periodicity detections (see Kovačević et al).

### 3.2. Structure Function Metric

Figure 2(a)-(d) present PDFs of SF-metric for each LSST OpSim cadence summed across the SF time scales  $\Delta t$  (left column of plots) and redshifts (right column of plots). The peak of each distribution shows the most likely metric value, i.e., square of deviation of SF based on LSST OpSim cadence from 'reference' SF based on ideal cadence. PDFs are marked with different colors for each LSST OpSim cadence, so that bluer corresponds to the LSST OpSim cadence with the smallest mean metric value (least deviation), while the redder color distinguishes the LSST OpSims cadences that have the immense mean metric value (highest deviations). The metric in r-band shows that SFs are more coherent with more minor deviations than those in g-band along the SF-time scale ( $\Delta t$ , compare plots (a) vs. (c)). Similar behavior can be seen across objects' redshifts (compare (b) vs. (d)). KL-divergence of SF metric in g vs. r-band is given in auxiliary Table 2 (https://github.com/LSST-sersag/white\_paper/blob/main/data/table\_2.pdf). Cadences, baseline\_samefilt\_v1.5\_10yrs and bulges\_cadence\_bs\_v1.5\_10yrs produce significant KL-divergence of metrics in g and r-band across redshifts. The DDF cadences are very likely to produce more reliable SFs (see Kovačević et al). The mean sampling of u-band cadences, such as baseline\_u\_ra\_0.0\_de\_- 10.0, baseline\_u\_ra\_0.0\_de\_- 30.0, rolling\_cadence\_0.8\_u\_ra\_0.0\_de\_- 10.0, rolling\_cadence\_0.8\_u\_ra\_0.0\_de\_- 30.0, are 58.2, 75.1, 64.8, 66.7 days respectively. Such condition limit the interpretation of the AGN time-lag and periodicity analysis.

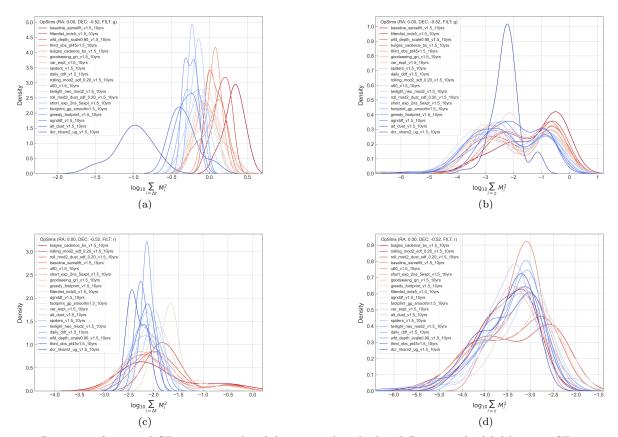


Figure 2. Densities of summed SF-metric in g-band (upper row) and r-band (lower row): (a),(c) across SF-time scale ( $\Delta t$ ) and (b), (d) redshifts. OpSim cadences are given in legends. Cooler colors correspond to the OpSim cadence with the smaller metric mean value (least deviation), while the warmer colors denote a larger metric mean value.

## 4. ANSWERS TO QUESTIONS

Q1: Are there any science drivers that would strongly argue for, or against, increasing the WFD footprint from 18,000 sq. deg. to 20,000 sq.deg.? Note that the resulting number of visits per pointing would drop by about 10%. If available, please mention specific simulated cadences, and specific metrics, that support your answer.

# Answer 1 Not applicable

Q2: Assuming that current system performance estimates will hold up, we plan to utilize the additional observing time (which may be as much as 10% of the survey observing time) for visits for the mini-surveys and the DDFs (with an implicit assumption that the main WFD survey meeting SRD requirements will always be the first priority). What is the best scientific use of this time? If available, please mention specific simulated cadences, and specific metrics, that support your answer.

Answer 2 Our detailed analysis in Kovačević et al suggest that DDF cadences are already very promising for determination of the reliable sizes of region sourounding supermassive black holes (and implicitly masses of these objects), detection of close-binary supermassive black holes candidates and reliable reconstruction of SF. In the long run, AGN variablity studies constraining AGN models and revealing nano-Hz gravitational wave candidates-will benefit if these DDF visits are enhanced rather than suppressed.

Q3: Are there any science drivers that would strongly argue for, or against, the proposal to change the u band exposure from 2x15 sec to 1x50 sec? If available, please mention specific simulated cadences, and specific metrics, that support your answer. Answer 3 Not applicable.

Q4: Are there any science drivers that would strongly argue for, or against, further changes in observing time allocation per band (e.g., skewed much more towards the blue or the red side of the spectrum)? If available, please mention specific simulated cadences, and specific metrics, that support your answer.

Answer 4 For size-determinations of regions surrounding AGN (and their masses consequently) our metric assumes that filters can catch both continuum and emission lines. However, different filters will cover different distances from continuum sources. In that sense we anticipated non-skewed time allocation per band.

Q5: Are there any science drivers that would strongly argue for, or against, obtained two visits in a pair in the same (or different) filter? Or the benefits or drawbacks of dedicating a portion of each night to obtaining a third (triplet) visit? If available, please mention specific simulated cadences, and specific metrics, that support your answer.

Answer 5 Reverberation-mapping requires simultaneous monitoring in different filters. In that sense, visits in a pair (or in triplet) in different filters will be of particular advantage.

Q6: Are there any science drivers that would strongly argue for, or against, the rolling cadence scenario? Or for or against varying the season length? Or for or against the AltSched N/S nightly pattern of visits? If available, please mention specific simulated cadences, and specific metrics, that support your answer.

Answer 6 We simulated AGN variability metric for rolling cadence scenario in Kovačević et al. From the OpSim 1.6 we used rolling cadences. Results based on the newest cadence release, OpSim 1.7, are given at (https://github.com/LSST-sersag/agn\_cadences). For example, OpSim\_ roll.cad\_0.8\_g\_RA\_0.0\_D\_-10.0 comprises 87 and OpSim\_ roll.cad\_0.8\_r\_RA\_0.0\_D\_-10.0 204 observations over 10 yr. The mean sampling of uband cadences, such as baseline\_u\_ra\_0.0\_de\_ - 10.0, baseline\_u\_ra\_0.0\_de\_ - 30.0, rolling\_cadence\_0.8\_u\_ra\_0.0\_de\_ - 10.0., rolling\_cadence\_0.8\_u\_ra\_0.0\_de\_ - 30.0, are 58.2, 75.1, 64.8, 66.7 days respectively. Such condition limit the interpretation of the AGN time-lag and periodicity analysis. Our metrics indicate that denser rolling cadences are more reliable for extraction of finer level of information such as AGN variability observables.

Q7: Are there any science drivers pushing for or against particular dithering patterns (either rotational dithers or translational dithers?) If available, please mention specific simulated cadences, and specific metrics, that support your answer. Answer 7 Not applicable.

## REFERENCES

Kovačević, A., Ilić, D., Popović, L. Č., Radović, V., Jankov, I., Yoon, I., Caplar, N., Čvorović-Hjadinjak, I., Simić, S., 2021, submitted to MNRAS

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