CONSTRANTS ON THE UBIQUITY OF CORONAL X-RAY CYCLES

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Received 2012 July 10; accepted 2012 September 21; published 2012 October 29

ABSTRACT

Stellar activity cycles are known to be a widespread phenomenon amongst moderately active solar- and late-type stars from long-term periodic variations in chromospheric Ca ii H and K emission lines, yet to date, only a handful of coronal X-ray cycles are known. We have surveyed serendipitously observed stellar sources in fields observed multiple times in the last decade by XMM-Newton and present our analysis of nine stars from six fields. Since our sample is flux-limited, it is strongly biased toward higher levels of X-ray activity. We fit a single temperature apec spectrum to each source and search for significant periodicities using a Lomb–Scargle periodogram. We use a Monte Carlo (MC) algorithm to yield robust analysis of the statistical significance of cycle detections and non-detections. None of the nine stellar light curves show any convincing indications of periodicity. From MC simulations, we simulate the detection capabilities of our methodology and, assuming a uniform distribution of cycle periods and strengths over the domain searched, we conclude with 95% confidence that less than 72% of the stars represented by our sample of active stars have 5–13 year coronal X-ray cycles.

Key words: stars: activity – stars: coronae – X-rays: stars

Online-only material: color figures

1. INTRODUCTION

The 11 year sunspot cycle was discovered over 160 years ago by Schwabe (1843). Later, the subsequent discovery by Hale (1908) that the global solar magnetic field reverses polarity after each sunspot cycle led to our current understanding of the 11 year solar activity cycle as arising out of an underlying 22 year magnetic cycle. In 1966, in order to ascertain whether or not other stars exhibited long-term periodic behavior, O. C. Wilson started a long-term monitoring program of chromospheric activity in solar-type main-sequence stars. He defined the S-index, which characterizes the strength of the Ca ii H and K lines relative to the photospheric continuum (Wilson 1968). Through monitoring a wide range of stars in the chromospheric Ca ii H and K emission lines via the S-index, we now understand that the activity cycle of our Sun is a more general phenomenon amongst solar-type stars (Wilson 1978; Baliunas et al. 1995). What factors determine the durations of these cycles, however, remains elusive to theoretical explanation.

The Ca ii H and K lines are well-established indicators of magnetic activity. Additionally, virtually all solar- and late-type main-sequence stars, including the Sun, have hot coronae that emit in the X-ray bands (Vaiana et al. 1981). The energy required to heat the corona is supplied by the stellar magnetic field, the dynamics of which are governed by an α–Ω dynamo involving differential rotation and turbulence in the convective envelope (e.g., Rosner et al. 1985; Ossendrijver 2003 and references therein), which is clearly manifested in the stellar activity–rotation relationship: stars that rotate faster are typically more X-ray luminous (e.g., Pallavicini et al. 1981; Wright et al. 2011).

Early efforts to statistically determine a trend of long-term variability in X-ray stars from Einstein and ROSAT data are summarized in Stern (1998). Stars that are X-ray faint typically show low levels of short-timescale variability (Feigelson et al. 2004; Wright et al. 2010), but may vary on longer timescales. Micela & Marino (2003) compare the solar variability to stellar data and find that this is consistent with cycles explaining much of the apparent luminosity spread in solar-like3 stars. Stars that are X-ray bright ($L_X > 10^{28}$ erg s$^{-1}$) are not believed to exhibit long-term X-ray cycles and instead have X-ray luminosities that vary erratically on short timescales (see review by Stern 1998). In this paper, we add experimental credence to this assertion by demonstrating an absence of long-term cycles in X-ray active solar-like stars.

Currently, long-term X-ray cycles have been found in only a handful of stellar sources: 61 Cygni A (Hempelmann et al. 2006), HD 81809 (Favata et al. 2008), α Centauri B (Robrade et al. 2005; Ayres et al. 2008; Ayres 2009), and our Sun. The amplitudes of the stellar X-ray cycles in these objects are typically larger than their respective S-index cycles by up to an order of magnitude, as is the case in the Sun (Judge et al. 2003).

The large levels of periodic X-ray variability expected from solar-like main-sequence stars mean that their cycles should be detectable with modern X-ray observatories such as XMM-Newton that have been in service for over a decade. In this article, we analyze serendipitous stellar sources detected in XMM-Newton fields that have been frequently observed over the lifespan of the instrument (~12 years) using publicly available data from the XMM-Newton archive. In Section 2, we discuss the observations used for this study and the methods employed. In Section 3, we present the observational results and Monte Carlo (MC) simulations to quantify our detection limits. In Section 4, we discuss the implications of these results on our understanding of stellar X-ray emission and long-term, solar-like X-ray cycles in other stars. We end with a presentation of our conclusions in Section 5.

* Based on observations obtained with XMM-Newton, an ESA science mission with instruments and contributions directly funded by ESA Member States and NASA.

3 “Solar-like” refers to spectral type (F-M), and not levels of X-ray activity.
Table 1: Field Characteristics

| Central Source | Field Name Abbreviation | Baseline (yr) | Distinct Observations |
|----------------|-------------------------|---------------|-----------------------|
| Capella        | CAP                     | 7.48          | 6                     |
| Zeta Puppis    | ZP                      | 9.98          | 13                    |
| PKS 2155−304   | PKS                     | 10.9          | 16                    |
| 3C 273         | 3C                      | 10.49         | 11                    |
| HD 81809       | HD                      | 6.0           | 5                     |
| Mkn 421        | MK                      | 10.98         | 7                     |

2. OBSERVATIONS AND DATA ANALYSIS

In XMM-Newton’s nearly 12 year history, numerous fields have been observed semi-regularly for purposes ranging from calibration (e.g., Capella) to measuring long-term periodic variability in X-ray luminosities (e.g., HD 81809). These fields offer a rare opportunity for the detection of long-term stellar X-ray cycles, which are believed to correlate strongly with chromospheric activity cycles. While central sources have been analyzed (e.g., HD 81809; Favata et al. 2008), serendipitous sources have not, to our knowledge, been checked for periodic behavior. Thus, for fields which have a long and consistent history of XMM-Newton observations, we check for periodicities in the light curves of serendipitous stellar sources.

2.1. Field Selection

Target fields that have been frequently observed (i.e., that have a baseline of at least six years and have been observed during at least five semesters between 1999 and 2012) were selected for analysis. We will show below that all candidate sources are relatively faint. To provide enough counts for a stable fit of the spectrum, we only use observations with a good time interval of ≥10 ks for one or more of the MOS1, MOS2, and PN cameras. Any data mode is acceptable. For instance, even though modes PPW2 and PPW4 have small central windows that do not read out parts of the field near the central source, these exposures still contain most serendipitous sources in the field. Six fields pass these selection criteria. Table 1 summarizes the observation timing and baselines for the six fields used in this paper. “Distinct observations” are the number of half-year intervals between 1999 and 2012 containing at least one usable observation.

2.2. Source Selection and Identification of Stars

X-ray sources are selected by cross-matching the 2XMMi Serendipitous Source Catalogue (Watson et al. 2009) with the optical NOMAD catalog (Zacharias et al. 2004), using a matching radius of 5″ consistent with the XMM-Newton point-spread function (PSF) for off-axis sources. For bright central sources, the 2XMMi catalog contains many spurious detections in the vicinity that are related to readout streaks in the original data. These readout streaks are manually removed from our target list.

We impose a minimum flux limit of 25 counts in a 10 ks observation, which equates to a maximum flux uncertainty of 20% from Poisson statistics. For an observation of 10 ks and 25 counts, this equates to a flux limit of \(1.4 \times 10^{-14} \text{ erg cm}^{-2} \text{s}^{-1}\) for a typical stellar observation and stellar properties (solar abundance, plasma temperature of 1 keV, and observations using the medium filter for the MOS).

A number of steps are taken to identify stellar X-ray sources. First, because stars have relatively soft X-ray spectra compared to most extragalactic sources, we require a hardness ratio of HR1 > 0 (see Watson et al. 2009). We then search the literature for available photometry and spectroscopy to allow us to classify sources as stellar or non-stellar. Due to the large number of non-stellar sources in deep X-ray observations, we require a high-confidence verification of the stellar nature of our candidates, discarding all sources that lack sufficient data to confidently classify them. We use three methods for identifying stellar sources based on the availability of certain observations and measurements, which, in order of priority, are as follows.

1. Spectra showing that the source is stellar in nature. Available for five targets (see Table 2).
2. Proper motions showing that the source is sufficiently nearby as to have a measurable motion on the plane of the sky with a significance of 3σ from the catalog of Röser et al. (2008). For stars with proper motion but no spectrum, we require that the photometry—if available—is also consistent with stellar-like colors (Covey et al. 2007). Three targets meet these criteria (sources 4, 5, and 7).
3. Photometry indicating stellar-like colors (Covey et al. 2007) in all bands. One target meets this criterion (source 3).

In addition to these requirements, stellar targets without available spectral information are required not to have been identified as extragalactic by any other studies incorporating observations at other wavelengths, morphology, or more complex multi-wavelength diagnostics. These criteria reduce the initial list of 72 sources to a high-confidence list of nine stellar sources (see Table 2).

2.3. Spectral Types and Distances

For these nine stellar sources, spectral types are determined from the best available data, either via spectroscopy (for five sources) or from the photometric color with the largest baseline (for four sources) and using the empirical data of Pecaut et al. (2012). All sources are assumed to lie on the main sequence, a reasonable assumption since our targeted fields are away from known star-forming regions and X-ray emission from late-type post-main-sequence stars is rare and weak (Linsky & Haisch 1979). The determined spectral types range from F-type to early M-type. Extinctions were calculated based either on the available photometry (for stars with known spectral types) or from an initial estimate of the distance (from the reddened magnitudes) and assuming an extinction of \(A_v = 1 \text{ mag kpc}^{-1}\), up to a maximum extinction of the total integrated Galactic extinction in that line of sight from Schlegel et al. (1998). Distances were determined using Two Micron All Sky Survey (2MASS; Skrutskie et al. 2006) \(K_s\)-band photometry and the empirical absolute magnitudes from Pecaut et al. (2012). For one source, Gliese 195 A, a parallax is available (Jenkins 1952) and is therefore used to determine its distance. The determined stellar distances range from 14 to 663 pc.

2.4. Data Analysis

Observational data are obtained from the XMM-Newton archive and analyzed using SAS (version 11.0.0; Gabriel et al. 2006). We apply all standard selection criteria for rejecting cosmic rays and filtering out the background rate to avoid contamination from solar proton flares. For all sources, spectra are extracted and the appropriate response files are generated.
Fitting of spectral models is performed in Sherpa (Freeman et al. 2011).

Due to the rather large PSF of XMM-Newton and the relatively dim nature of the selected sources, a 450–500 pixel (∼8′–9′) radius is used for the source region and a 2500 pixel (∼46′) radius is used for the background region, manually chosen to avoid other 2XMMi sources. One background region is chosen for each field to simplify the processing.4

2.5. Fitting apec Models

A single temperature apec model is fitted to extracted spectra that contain a minimum of 125 counts within the 0.4–7 keV range after subtracting background counts. We fix the interstellar absorption using the $A_V$ values from Table 2 and the relation $N_H = 1.8 \times 10^{21} (1/\text{mag cm}^2) A_V$ (Predehl & Schmitt 1995). We employ the Levenberg–Marquardt (LM) fitting algorithm (Marquardt 1963, called lmfit in sherpa) with a Gehrels’s $\chi^2$ statistic (Gehrels 1986) to fit apec models to the observed spectra. If the Gehrels’s $\chi^2$ value of an LM fit exceeds a threshold value heuristically set to 0.7, then we seek an improvement of $\chi^2$. First, the LM algorithm is again tried with five arbitrary and distinct initial parameter values. If the Gehrels’s $\chi^2$ of the fit is still above the heuristic threshold, then the fitting method is changed to the more accurate but slower differential evolution method (Storn & Price 1997; called moncar in sherpa) as a second effort for improving the fit.

At the end of the fitting procedure, if the Gehrels’s $\chi^2$ is below 0.25 or above 3, $1\sigma$ flux uncertainties are not computed. A $\chi^2$ above three strongly suggests that the model does not adequately fit the data. A $\chi^2$ below ∼0.25 suggests that the apec model fit is not well constrained. Flux values without estimated $1\sigma$ uncertainties are not used in later analysis.

Unless otherwise stated, X-rays are defined to be in the 0.4–7 keV band. $F_X$ is the integrated energy flux over this range. $L_X$ is the unabsorbed luminosity over this range. The $F_X$ values are obtained from integrating the fitted apec model. Fit errors for the integrated $F_X$ are calculated from the uncertainty of the model normalization.

3. RESULTS

3.1. The Stellar Sample

Table 2 summarizes the properties of our stellar sample. Uncertainties in $L_X$ and $L_X/L_{bol}$ are about 0.2 dex, including the statistical uncertainty from the fitted X-ray model and the uncertainty in distance from the photometry. The spectral types of our sample range from F to early M, and the activity level (log($L_X/L_{bol}$) from −3 to −6. Active stars are known to saturate at an activity level of log($L_X/L_{bol}$) ∼ −3, while the activity level drops for older or more slowly rotating stars (Stauffer et al. 1994). Naturally, stars as inactive as our Sun are not part of this serendipitous sample because they are too X-ray faint.

3.2. The Lomb–Scargle Periodogram

After fitting a single temperature apec model to the spectra of all nine stellar sources and integrating the fitted model over the 0.4–7 keV range to obtain flux estimates for each exposure, the resulting light curve is analyzed for the presence of periodic behavior. This is done using a Lomb–Scargle periodogram (LSP; Lomb 1976; Scargle 1982). Derived quantities from the LSP are (1) the frequency of the best-fit sinusoidal function to the data (known as the peak frequency) and (2) the “false alarm probability” (FAP) of the strongest periodicity found. The FAP is defined to be the probability that periodic behavior of a certain strength could be found in Gaussian noise.

3.3. Cycle Detection

Baliunas et al. (1995) consider an FAP ≤ 0.001 (≥99.9% confidence) as a detection of periodicity in their analysis of S-index data from the Mount Wilson Observatory. In our case, X-ray flux uncertainties are large and most stars have been observed sparsely in the last 12 years. Thus, we do not expect the light curves of the stellar sources in this paper to produce such high-confidence statistics.

Instead of employing a simple FAP cutoff, we use an MC algorithm to assess the statistical likelihood of LSP results by taking into consideration the flux uncertainties, which vary from observations to observation. We consider a light curve to be periodic (or at least warrant further investigation) if 68% of all MC-sampled light curves have an FAP below 0.30.

Table 2

| No. | Field Name | R.A. (h:m:s) | Decl. (d:m:s) | Spectral Type | $V$ (mag) | $K_s$ (mag) | $d$ (pc) | log($L_X$) (erg s$^{-1}$) | log($L_X/L_{bol}$) |
|-----|------------|-------------|---------------|---------------|----------|------------|--------|-----------------|------------------|
| 1a  | CAP        | 05:16:32.0  | 46:08:27.6    | F0           | 8.06     | 7.23       | 0.11   | 123             | 28.6             | −5.8             |
| 2a  | CAP        | 05:17:24.0  | 45:50:20.4    | M1           | 10.16    | 5.95       | 0.13   | 13.9            | 28.2             | −4.0             |
| 3   | ZF         | 08:04:22.6  | −39:50:28.7   | G0           | 13.37    | 11.57      | 0.43   | 419             | 30.3             | −3.6             |
| 4   | HD         | 09:27:12.3  | −05:57:16.9   | K3           | 15.46    | 13.07      | 0.11   | 572             | 30.5             | −2.6             |
| 5   | HD         | 09:28:34.3  | −06:03:45.7   | F9           | 13.37    | 11.91      | 0.11   | 663             | 30.3             | −3.9             |
| 6a  | MK         | 11:04:43.8  | 38:14:48.5    | E5           | 7.51     | 6.39       | 0.05   | 64.9            | 29.0             | −5.1             |
| 7   | 3C         | 12:28:28.6  | 01:54:49.3    | K0           | 12.68    | 10.67      | 0.06   | 240             | 29.0             | −4.4             |
| 8   | 3C         | 12:28:37.2  | 01:57:20.9    | M1h          | 14.09    | 10.04      | 0.06   | 79.5            | 28.8             | −3.4             |
| 9   | 3C         | 12:29:42.4  | 01:55:25.3    | G0           | 12.88    | 10.89      | 0.06   | 383             | 30.3             | −3.5             |

Notes. Spectral types extracted from the literature where a citation is provided or from available colors. $V$ magnitudes taken from Høg et al. (2000) and the NOMAD catalog (Zacharias et al. 2004). $K_s$ magnitudes extracted from the 2MASS catalog (Skrutskie et al. 2006) and distances determined photometrically from this (with the exception of source 2, which has a parallax).

Other names: aBD+45 1076; bCapella H, G 96-29, Gliese 195 A; cHD 95976.

References. 4Heckman 1975; 5Montes et al. 2001; 6Cannon & Pickering 1918; 7López-Santiago et al. 2007; 8Covey et al. 2008; 9Jenkins 1952.
3.4. Monte Carlo Simulations

While periodogram analysis offers an invaluable statistical method for determining the presence of periodicities, the LSP calculation itself does not account for uncertainties in the data. Therefore, we use an MC algorithm to determine the statistical significance of a detected periodicity.

In a light curve containing \( T \) data points with uncertainties, the MC algorithm samples \( T \) flux values in the following way for each step (\( N = 1000 \) total steps) of the simulation:

\[
F_{MC}(t_i) = F_i + X_i ,
\]

where \( F_i \) is the \( i \)th observed flux (at time \( t_i \)) and \( X_i \) is a random variable sampled from a Gaussian distribution of mean zero and a standard deviation equal to the uncertainty of the empirical flux value. Figure 1 shows the light curve, a cumulative histogram of FAP values, and a histogram of the strongest periods (inverse of the peak frequency in the LSP) detected in star 7, a K1 star in the 3C 273 field.

3.5. The Test Case: HD 81809

As a test case, we present an analysis of a stellar source with a known X-ray cycle. HD 81809 has an 8.2 year \( S \)-index cycle (Baliunas et al. 1995) and an X-ray cycle that appears in phase with the \( S \)-index cycle (Favata et al. 2008). We use all publicly available \textit{XMM-Newton} data for HD 81809 and carefully determine the presence and significance of any periodicities in its light curve. Figure 2 displays fitted \texttt{apec} flux values as well as an LSP. Indeed, the periodic behavior has continued with what appears to be an \( \sim 8 \) year period.

The LSP produces a local minimum at the expected period surrounded by two peaks at a period of 6.2 years and 12 years. The MC simulations confirm that the 6 yr peak has an FAP of \(<5\%\) in 100\% of the MC steps. The MC simulations also show that the shift in period compared to the \( S \)-index observations is an artifact caused by infrequent time-sampling of the \textit{XMM-Newton} observations compared to the observations of Baliunas et al. (1995). Similar behavior can be found in the Fourier analysis of X-ray data for 61 Cyg, where the double peak in the Fourier power close to the known \( S \)-index period is revealed to be an artifact of infrequent and semi-regular observation times: a similar phenomenon occurs in the Fourier analysis of \( S \)-index data when sampled at the same times as X-ray observations (Hempelmann et al. 2006).

3.6. Detection Capabilities

Despite the inaccuracy of the exact cycle period, analysis of HD 81809 shows that our method for determining the presence of stellar cycles is statistically sound. No significant periodic behavior was found in the light curves of any of the nine stars in our sample.

To determine which cycles our method is capable of detecting, we use MC simulations again, assuming that a source’s X-ray flux over time is

\[
f(t) = F_{avg} + F_{amp} \sin \left(\frac{2\pi t}{P} + \delta \right) \quad (1)
\]
and that the observations are subject to Gaussian uncertainty with standard deviation, \( \sigma \). Thus, for a given period \( P \) and cycle amplitude relative to the average flux \( (F_{\text{amp}}/F_{\text{avg}}) = s \), an MC simulation creates \( T \) randomly spaced data points over a span of \( S \) years, produces an LSP, and calculates the FAP of the best-fit sine function. \( N \) sampled light curves are created for each \((P, s)\) value. We then calculate the fraction of \( N = 1000 \) MC light curves that produce an FAP less than 0.3. Several contour plots are shown in Figure 3.

Contour plots like those in Figure 3 are calculated via a similar MC algorithm for each individual source. Instead of sampling \( N \) observations randomly distributed over the baseline with a constant uncertainty, the observation dates and flux uncertainties for the particular star of interest are used to generate sample light curves in the MC algorithm.

4. DISCUSSION

The time span covered by the observations differs for each analyzed field and ranges between 6 and 11 years. Most sources have more than 30 exposures with a usable time longer than 10 ks. However, several exposures are often taken within a few days and thus do not contribute more information to the periodogram. Typically, about 10 data points with sufficient spacing exist. The statistical errors depend on the source flux and the exposure time. Source flux varies from source to source, and exposure time varies even between exposures. Here, a typical value of relative error on the flux is about 25%.

We now seek to use our results to put an upper limit on the prevalence of coronal X-ray cycles in stars of spectral type F through M that have 5–13 year periods. We calculate \( P(\text{FAP} < 0.3) > 0.68 \) contours for each source and use these to create a binary function over the \( P-s \) domain that is equal to 1 when \( P(\text{FAP} < 0.3) > 0.68 \) and 0 otherwise. The binary functions for each star are summed to generate a map \((Z(P,s))\) that represents the number of cycles that could have been detected using our method for a cycle of particular strength and period.

We assume for simplicity that the intrinsic rate of coronal X-ray cycles is approximately constant across spectral types F–M. This assumption may or may not be justified: the S-index monitoring presented in Baliunas et al. (1995) did not investigate M stars (except for one for which no cycle was found); however, monitoring by Mauas et al. (2012) has detected possible S-index cycles in M dwarfs that fall within our 5–13 year domain. Baliunas et al. (1995) do see differences in chromospheric cy-
cycles between F-, G-, and K-type stars, such that, e.g., almost all K-type stars with a low S-index have a cycle, whereas F-type stars with low S-index typically have flat activity curves. However, the weakness of the chromosphere and the low contrast of the photosphere make the detection of cycles on F-type stars more difficult, so that it is not clear if the cycle frequency actually varies with spectral type.

For a particular kind of cycle \((P_s, s)\), \(Z(P_s, s)\) is then the effective sample size. The probability of not detecting a \((P_s, s)\)-type cycle is \((1 - R)^{Z(P_s, s)}\), where \(R\) is the intrinsic rate of \((P_s, s)\)-type coronal X-ray cycles for F–M stars. A 2\(\sigma\) upper limit on \(R\) would imply that \(R \leq R_{2\sigma}\), where \(R_{2\sigma}\) is defined to be such that the probability of not detecting a certain type of cycle, \((1 - R_{2\sigma})^{Z(P_s, s)}\), is 0.05. Solving for \(R_{2\sigma}\) in terms of \(Z\), \(R_{2\sigma}(P, s) = 1 - (0.05)^{1/Z}\). A plot of \(Z\) is provided in Figure 4, with implied \(R_{2\sigma}\) values provided in parentheses.

While there is a notable area where \(R_0\) is high (i.e., where the 2\(\sigma\) upper limits on the underlying rate of coronal X-ray cycles are weak), the statistics from this sample strongly limit the prevalence of short-term cycles (five to eight year periods) in our sample. Since both the Sun and HD 81809 have \(s\) values close to 0.7, we can say with 95% confidence that fewer than \(\sim 50\%\) of the stars in our sample have five to eight year X-ray cycles with amplitudes of \(s = 0.7\).

In addition, if one makes the assumption that coronal X-ray cycles should have strengths and periods that are more or less uniformly distributed over \((0 \leq P \leq 1)\) and \((5 \text{ years} \leq P \leq 13\text{ years})\) domain, then Figure 4 implies a 2\(\sigma\) upper limit on the ubiquity of coronal X-ray cycles in stars represented by our sample. Under this assumption, our statistics imply that less than \(\sim 72\%\) of those stars have coronal X-ray cycles.

All statistical constraints given thus far are valid for a stellar population that is well represented by our sample, i.e., the population of stars detected in a serendipitous X-ray survey. Since we are limited by the flux of sources in the field, our sample is biased toward X-ray active stars (Malmquist-bias). Only two sources have activity levels \(L_X/L_{bol} < -4.5\) and even the least active one is still more active than the Sun by an order of magnitude or more. Thus, we cannot draw reliable conclusions about F–M stars in general from our data. This requires a volume-limited sample as, e.g., in Schmitt & Liebke (2004).

The lack of detection of activity cycles in our sample would therefore support the view that in active stars, coronal X-ray cycles are not common. This was observed in the S-index monitoring by Baliunas et al. (1995). In X-rays, the absence of cycles in active stars has been suggested by Stern (1998) on the bases of more fragmentary data. As stars age, their median luminosity decreases (see reviews of Rosner et al. 1985; Favata & Micela 2003 and references therein) and they develop coronal cycles with larger amplitudes of variation (Micela & Marino 2003).

5. CONCLUSIONS

We have identified nine stellar X-ray sources that have frequently been observed in repeat XMM-Newton observations over the lifetime of the observatory. From single temperature apec model fits to extracted spectra, we have generated light curves of all stellar sources and tested the light curves of these stars for the presence of underlying periodicities using an LSP assisted by an MC algorithm. No periodicities were found.

Using MC simulations, we calculated which kinds of cycles should be detectable with our methodology for each of the nine stars in our sample. We term the underlying rate of coronal X-ray cycles in F–M stars that have 5–13 year periods as \(R\). We assume that \(R\) is constant across spectral types F through M and that the short-term X-ray variability of our stellar sources is negligible compared to any long-term periodic behavior. Using these assumptions, we calculate 2\(\sigma\) upper limits on \(R\) for each type of cycle investigated in this study (cycles of any strength that have 5–13 year periods). Our sample is biased toward stars with higher levels of X-ray activity; the least active star is at least an order of magnitude more X-ray active than our Sun. If one makes the assumption that X-ray active solar-type stars have cycles with periods and strengths that are uniformly distributed over the \((0 \leq s \leq 1)\) and \((5 \text{ years} \leq P \leq 13\text{ years})\) domain, then we can conclude with 95% confidence that less than 72% of the active main-sequence stars have 5–13 year coronal X-ray cycles of any strength. We stress that these numbers do not apply to the stellar population as a whole, which contains a large fraction of stars that have significantly lower levels of X-ray activity.

J.H. thanks Marie Machacek and Jonathan McDowell for their invaluable guidance and assistance with this project throughout the summer of 2011. This work is supported in part by the National Science Foundation Research Experiences for Undergraduates (REU) and Department of Defense Awards to Stimulate and Support Undergraduate Research Experiences (ASSURE) programs under grant No. 0754568 and by the Smithsonian Institution. We thank an anonymous referee for help in improving the paper.

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