New BRITE-Constellation* observations of the roAp star α Cir

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ABSTRACT

Context. Chemically peculiar (CP) stars with a measurable magnetic field comprise the group of mCP stars. The pulsating members define the subgroup of rapidly oscillating Ap (roAp) stars, of which α Cir is the brightest member. Hence, α Cir allows the application of challenging techniques, such as interferometry, very high temporal and spectral resolution photometry, and spectroscopy in a wide wavelength range, that have the potential to provide unique information about the structure and evolution of a star.

Aims. Based on new photometry from BRITE-Constellation, obtained with blue and red filters, and on photometry from WIRE, SMEI, and “we attempt to determine the surface spot structure of α Cir and investigate pulsation frequencies.

Methods. We used photometric surface imaging and frequency analyses and Bayesian techniques in order to quantitatively compare the probability of different models.

Results. BRITE-Constellation photometry obtained from 2014 to 2016 is put in the context of space photometry obtained by WIRE, SMEI, and “This provides improvements in the determination of the rotation period and surface features (three spots detected and a fourth one indicated). The main pulsation frequencies indicate two consecutive radial modes and one intermediate dipolar mode. Advantages and problems of the applied Bayesian technique are discussed.

Key words. Space photometry – Stars: chemically peculiar – Starspots – Pulsation – Rotation – Stars: individual HD 128898, α Cir

1. Introduction

Rapidly oscillating chemically peculiar (roAp) stars are a subgroup of the chemically peculiar (CP) stars with spectral types ranging from late B to early F, with luminosity class IV to V, and a global magnetic field that causes a peculiar chemical composition of their atmosphere and that deviates from solar composition (Kurtz1990). The roAp stars provide important tools for investigating stellar structure and evolution and for testing astrophysical concepts. These stars show low-degree, high-order acoustic oscillations, similar to the 5 min oscillations in the Sun, but with coupling to the magnetic field. The excitation of such high-order oscillations, instead of the low-order pulsation in δ Scuti stars, is expected to be related to the magnetic field, which suppresses the near-surface convection, and therefore increases the efficiency of the opacity mechanism in the hydrogen ionisation region (Balmforth et al. 2001). In principle, the high-order oscillations should simplify the theoretical interpretation of the observed power spectrum by the use of standard asymptotic tools, but since the oscillations are modified near the surface by the magnetic field, more sophisticated modelling is required (Cunha 2006). However, the coupling between rotation, pulsation, and magnetic field in a chemically peculiar environment makes the roAp stars important laboratories for testing stellar structure and evolution theory.

The star α Cir is visible to the naked eye (V = 3.2 mag). It is a well-known and well-observed member of the roAp group, and has a mean quadratic magnetic field of 4 kG to 7 kG (Mathys 2006), a main pulsation period of 6.8 min, and a rotation period of about $P_{\text{rot}} = 4.48$ d. Weiss et al. (2016) hereafter Paper I have already summarised past investigations of α Cir, provided references, and discussed striking differences between the rotation light curves deduced from red and blue filter BRITE-Constellation observations (Weiss et al. 2014).

The data needed for clarification were obtained by three of the five BRITE-Constellation satellites in 2016. In addition we

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analysed WIRE data from 2006 and 2007 (Bruntt et al. 2009), and data from SMEI obtained from 2003 to 2010 (Jackson et al. 2004; Tarrant et al. 2008) which cover the two last WIRE seasons. Finally, we also address "data (Ricker et al. 2015) obtained in 2019.

The large time base (almost 20 years of observing α Cir with various space missions) allows us to discuss rotation and pulsation, with a discussion of the inherent problems of photometric surface imaging using Bayesian techniques, which has two main advantages: (i) parameter estimation with the Bayesian technique provides credibility regions from the data alone and (ii) it allows us to rank models according to their evidence.

Concerning pulsation properties, the dependence of amplitude and phase of the dominant frequency $f_1$ on the rotation phase of α Cir (see Paper I) was another challenge for the 2016 BRITE photometry. Merging the results with space photometry obtained by WIRE, SMEI, and "helped to improve the accuracy significantly. With a much better signal-to-noise ratio (S/N) of the new red and blue 2016 BRITE data, we also address other frequencies mentioned by Bruntt et al. (2009) and in Paper I.

### 2. Photometric data and reduction

As α Cir was observed by different satellites at different epochs, the various data sets exhibit individual peculiarities. The effective wavelengths of the passbands of the various satellites (Table 1) were determined by using the filter values and a synthetic spectrum with $T_{\text{eff}} = 7500 K$, $\log g = 4.1$ [cm s$^{-2}$], and a chemical composition that agree with the atmospheric parameters determined by Kochukhov et al. (2009) for α Cir.

| B & BLb (= Bb*) | 425 | 55 |
|-----------------|-----|----|
| BTr             | 605 | 145|
| WIRE            | ≈600| ≈380|
| SMEI            | 630 | ≈600|
| TESS            | ≈800| ≈400|

#### 2.1. BRITE-Constellation

α Cir was observed during the commissioning phase of BRITE-Constellation (Weiss et al. 2014) from March 3, to August 8, 2014, for 146 days. The analysis of these data is presented in Paper I, which describes a first attempt at photometric surface imaging of α Cir in blue and red colours, based on Bayesian techniques and using the rotation period of 4.4790 days, determined by Kurtz et al. (1994).

The star was observed again in the 15-CruCar-I-2016 field from February 4, to July 22, 2016, by three of the five operational nanosatellites, which provided a total of 163 433 photometric measurements from BTr, BAb, and BLb (Table 2) and which are the basis for the present investigation. Each satellite obtained 10 to 30 measurements per satellite orbit (about 101 min) with a typical sampling of 20 s and exposure time of 1 s.

The pipeline outlined by Popowicz (2016) and Popowicz et al. (2017) was used to process the raw images from the satellites. While the pipeline accounts for technical issues typical of the BRITE photometry, such as hot pixels (Pablo et al. 2016), the extracted photometry remains affected by systematic instrumental effects. These systematics, such as CCD temperature drifts and the position of the star’s point spread function (PSF) in the raster, are identified and removed via decorrelation using a procedure similar to that outlined by Pigulski et al. (2016) and Pigulski (2018). An additional decorrelation following the procedure outlined by Buysschaert et al. (2017) mitigates the impact of the PSF modulation with temperature.

#### 2.2. WIRE

The Wide Field Infrared Explorer (WIRE) was launched on March 4, 1999, but the hydrogen cryogen boiled off soon after launch due to a technical problem, which terminated the primary science mission. The onboard star tracker with 52 mm aperture, however, remained functional and could be used for long-term visual precision photometry until communication with the satellite failed on October 23, 2006 (Buzasi 2002; Brunnt 2007; Brunnt & Southworth 2008). Our target, α Cir, was observed in September 2000; February 2005, and February and July 2006, for a total of 84 days (Brunnt et al. 2009), of which we are addressing the 2005 and 2006 data (February 16, 2005, to August 1, 2006).

#### 2.3. SMEI

The Solar Mass Ejection Imager (SMEI) is an instrument on board the Coriolis satellite, which was launched on January 1, 2003. The primary science goal is to detect disturbances in the solar wind, but in doing this the three CCD cameras observed the whole sky in successive passes. These data were used to detect, among other things, stellar pulsation (Jackson et al. 2004; Tarrant et al. 2008; Hounsell et al. 2010). Our target star α Cir was observed by SMEI from February 3, 2003, to December 30, 2010.

Raw SMEI data suffer from very strong instrumental effects. The final light curve was obtained by correcting first for a one-year periodicity, then detrending and sigma clipping, which was repeated up to 25 times. Finally, signals in the vicinity of 1, 2, ... 6 d$^{-1}$ were subtracted. The frequency spectrum of the final data has a sharp decrease at the lowest frequencies, which is the result of detrending. A signal at the satellite orbital frequency, its multiples, and side lobes at typically ±1 and ±2 d$^{-1}$ were also removed. Unfortunately, we do not know all the details of the SMEI photometry. A single SMEI data point comes from a series of 4 s exposures, but it is not always clear how many individual exposures are combined. This depends on the time a star passes through the camera field of view (3° × 60°), and changes with the aspect angle. Thus, the total integration time is typically below one minute, but can sometimes be slightly longer.
2.4. TESS

TESS was launched on April 18, 2018, and has four identical wide-field cameras that together monitor a 24° by 96° strip of the sky and with a red-optical passband. Each field is monitored for about 27 days.

The TESS data for α Cir span from April 24, to June 18, 2019, in sectors 11 and 12 with a baseline of 55 days (four orbits of 13.7 days each) with a 2 min cadence. We downloaded the Simple Aperture Photometry from the MAST portal[1] applied a 3σ outlier clipping (relative to a 2-day moving average), and used the data without any further corrections.

3. Bayesian photometric imaging

Our Bayesian photometric imaging technique is described in detail by [Lüftinger et al. 2010] and in our Paper I. The number of free parameters $N$ depends on the complexity of the stellar surface model considered. The following spot parameters are estimated: longitude, latitude, and radius for each spot, which is assumed to be circular. Therefore, a three-spot model involves at least ten free parameters, including the rotation period.

Published values were used for the following parameters:

- Inclination ($i = 36°$). This value results from $v \sin i$ and the stellar radius given by [Bruntt et al. 2008].
- Quadratic limb darkening ($U_a = 0.278, U_b = 0.382$). Limb darkening depends on wavelength, and also on rotation phase because α Cir is a mCP star. We chose representative values based on a model atmosphere with a $T_{\text{eff}} = 7500$ K and $g = 4.1$ (cm s$^{-2}$) [Kochukhov et al. 2009]. We note that the exact choice of the limb darkening coefficient is not critical here as it is degenerate with the spot size resulting from the light curve modelling.
- Contrast between spot and undisturbed photosphere. This has to be fixed because spot area and brightness contrast are highly anti-correlated. The spots are almost certainly bright in the optical wavelengths as the flux absorbed in the UV is redistributed to the red part of the spectrum for $T_{\text{eff}}$ values typical for mCP stars (see Section 4.2, of Paper I). To start we assume a contrast of $\kappa = 1.25$.

For our Bayesian surface imaging we use the average magnitude during each orbit of the respective BRITE satellite after carefully removing instrumental effects. This binning ensures that the brightness variation due to stellar oscillations (less than 2 mmag in the blue and even less in the red filter, and of the order of few minutes) cancels out. The much higher noise level of the BRITE-blue data[2] (BAb and BLb) compared to the red data (BTr) is evident. Fortunately, the amplitudes are larger in the blue.

Similarly, the 36 124 TESS data points were binned for our surface imaging into 4019 time bins, with a maximum bin size of 16.1 minutes. This results in up to nine original data points per bin with a typical scatter of ±0.37 mmag per data point. The accuracy of a bin-mean is up to three times better.

3.1. Bayesian concept

In a nutshell, a Bayesian approach consists of the following: An uninformative prior probability distribution in the N-dimensional parameter space is converted by the likelihood of the data, given a set of parameter values, into a posterior probability distribution. This posterior contains what can be learned from the data in terms of a given model. It is common practice to extract N marginal distributions, one for each parameter, from the posterior. Each marginal distribution can be summarised by a mean value and a 90% credibility interval. In a few cases we use a 68% interval.

In a second step, a model’s evidence can be determined by integrating the posterior probability distribution over the parameter space and dividing by the volume of the latter because it is the mean probability that is important. This is computationally demanding, but allows one to quantitatively compare different models (e.g. a two-spot model versus a three-spot model).

The centre of gravity (the barycentre of the N-dimensional posterior ‘probability mountain’) is defined by the N parameter mean (expectation) values, computed as described above; it minimises the mean quadratic dispersion. It is this barycentre that makes best use of all the available information as less probable sets of parameter values also contribute to a model’s evidence. This mean is more relevant than the locus of maximum probability, the mode (best fit). Only in the rare case of a symmetric probability mountain do mean and mode coincide.

A corresponding picture could be a comparison of the height of the Matterhorn in Switzerland with that of Table Mountain in Cape Town, South Africa, and quote their positions. The case of the Matterhorn is trivial. It has a nearly symmetric structure and an obvious peak above the timberline, but Table Mountain raises a problem. Is the position of the mini-peak at the border of the Table (location of the ‘mode’) to be listed, or is it more appropriate to quote the geographical centre (centroid, ‘mean’) of the plateau?

Probabilistic methods do not primarily aim for a single best-fit solution of a given problem, but for the posterior distribution of the various model parameters and how well a model reproduces the observations compared to other models. However, for the sake of completeness we also provide modal values in Table 3 which summarises the model parameters derived from BTr, WIRE, SMEI, and 4 observations.

![Fig. 1. Left: Spot 1 marginal distributions of longitude ($\lambda$) and latitude ($\beta$) in degrees. Right: Spot radius ($r$) and period in days. All for BTr data. The vertical lines indicate the mean and the 90% credibility limits (shaded area). Because of the skewness of the distribution (curved line) the locus of maximum probability does not necessarily coincide with a parameter’s mean value and deviates from the global mode given in Table 3 for BTr.](https://example.com/figure1.png)
Table 3. Estimated parameter values for spot models based on BTr, WIRE, SMEI, and TESS data.

| Spot   | BTr  | WIRE | SMEI | TESS |
|--------|------|------|------|------|
|        | mean | mode | mean | mode | mean | mode | mean | mode |
| spot 1: |      |      |      |      |      |      |      |      |
| longitude (°) | 0°±4 | 9 | 0°±2 | 0 | 0°±2 | 0 | 358 | 0.0±0.3 | 0.0 |
| latitude (°)   | -27°±14 | -63 | 37°±5 | 43 | -19°±16 | -10 | -2.7±0.6 | -2.4 |
| radius (°)     | 21°±12 | 52 | 14°±3 | 17 | 19°±20 | 14 | 9.7±0.1 | 9.7 |
| spot 2:      |      |      |      |      |      |      |      |      |
| longitude (°) | 171°±4 | 183 | 179°±4 | 178 | 155°±7 | 155 | 144.4±0.5 | 144.5 |
| latitude (°)  | -30°±12 | -37 | 40°±6 | 44 | 46°±25 | 67 | 20.7±0.6 | 20.9 |
| radius (°)    | 21°±9 | 24 | 13°±3 | 17 | 8°±2 | 10 | 6.7±0.1 | 6.7 |
| spot 3:      |      |      |      |      |      |      |      |      |
| longitude (°) | 111°±20 | 133 | 120°±2 | 120 | - | - | 171.9±0.6 | 171.7 |
| latitude (°)  | 34°±22 | -4 | -10°±2 | -11 | - | - | -36.1±5.1 | -36.8 |
| radius (°)    | 6°±1 | 9 | 9°±1 | 9 | - | - | 16.9±4.2 | 17.5 |
| period (days) | 4.4779 | 4.4781 | 4.47925 | 4.47926 | 4.47912 | 4.47912 | 4.4803 | 4.4803 |
| residuals (mmag) | ±0.0012 | ±0.00009 | ±0.00018 | ±0.0004 | ±0.0004 | ±0.141 |

Notes. Longitudes are with respect to HJD 2457510.5105 (BRITe-Toronto), HJD 2453680.6598 (WIRE), HJD 2454115.2887 (SMEI), and HJD 2458625.8020 (TESS). Spots are ordered according to their impact on the light curve. Modal values are provided together with mean values and 90% credibility limits. The set of mean parameter values may differ markedly from the set of modal values due to the skewness of the posterior. The residuals in mmag of the photometric data to the model light curve are r.m.s. values.

Figure 1 deduced from the BTr data, shows the marginal distributions for longitude, latitude, and radius of spot 1, which has the largest impact on the light curve, and the marginal distribution for the rotational period. Each marginal distribution results from integrating the N-dimensional posterior over the corresponding (N – 1)-dimensional subspace. The boundaries of the 90% credibility interval as well as the parameter mean are indicated by vertical lines. The widths and asymmetries of the marginal distribution for latitude and radius indicate that these two model parameters are poorly defined, whereas spot longitude and rotation period are much better constrained by the data. Because of the skewness of the distribution, the locus of maximum probability does not necessarily coincide with a parameter’s mean value and may be very different from the global mode presented in Figure 1.

It should be noted that finding the posterior’s mode is computationally fast in comparison to the Markov chain Monte Carlo (MCMC) integrations, which are necessary to provide mean values and credibility intervals.

3.2. Rotation

An early estimate for a rotation period, based on magnetic field measurements (Wood & Campusano 1975) and $H_a$ polarimetry (Landstreet 1980) is given by Mathys (1991), who speculated that the star has a magnetic field that varies with a period of more than two weeks. Various photometric, spectroscopic, and magnetic field investigations of α Cir were published in the following years claiming a rotation period between 4.46 d and 4.48 d (Kurtz et al. 1994), Balona & Laney (2003), Bychkov et al. (2005), Bruntt et al. (2009), Hubrig et al. (2004).

Figure 2 shows the rotational light curves of α Cir observed by six satellites. The differences between the shapes of the blue (Bb) and red light curves indicate a complex balance between line or continuum opacities, which removes flux from the blue band, but increases flux at other wavelength regions due to back-warming, as is discussed by Leckrone (1976) and Shulyak et al. (2010), among others. In particular, surface spots with different chemical compositions, typical of CP stars, can cause complex effects on light curves obtained with different filters. Extended spectroscopic investigations are needed to explore such a scenario.

The stellar rotation period is observationally the most obvious parameter and can be estimated, in principle, without a model. However, here we consider the rotation period as an unknown parameter of a three-spot model (Table 3). A re-analysis of the BRITe-red 2014 data from Paper I, with the rotation period being a further free parameter in the Bayesian analysis, indicates a $P_{rot} = 4.4846 ± 0.0017$ d (68% interval), which exceeds the standard value of 4.4790 d by 3.3σ.

A comparison of spot 1 transit times from WIRE and TESS (three-spot model) resulted formally in $P_{rot} = 4.47930 ± 0.00002$ d (68% interval), assuming that there are 1104 stellar revolutions between the corresponding maps (Fig. 3).
3.3. Spots

A major challenge is estimating the number of spots necessary to get a good fit without overfitting. In Paper I we were able to identify one bright spot in the blue data, two spots in red and WIRE data, but there were already hints of a more complex surface map, at least in the red case. Not surprisingly, the photometric quality of the data matters. In contrast to the 2014 data, the 2016 BTr data convincingly indicate a three-spot model. The reduction in credibility, which is a formal consequence of an increased number of free parameters from seven (two spots) to ten (three spots) is more than compensated for by an improved goodness-of-fit value.

With the MCMC technique and the availability of PC clusters, an integration over a high-dimensional configuration space is meanwhile computationally feasible. With computed evidence it is possible to quantitatively compare the performance of models differing in the number of spots by considering the evidence ratio (i.e. the Bayes factor). To our knowledge, this is the first time that the evidence of models differing in the number of spots has been probabilistically determined. The poor quality of the blue photometry prevented an independent estimate of $P_{\text{rot}}$ as a Bayesian spot modelling parameter.

3.3.1. Red maps

For the red data obtained with BTr, the evidence gain from a two-spot model to a three-spot model is approximately $1.2 \times 10^6$. This large Bayes factor is due to the number of data points (1678) involved. The residuals to the photometry decreased slightly from 1.389 to 1.386 mmag and the mean gain of evidence per data point is 0.8% ($(1.2 \times 10^6)^{1/1678} = 1.008$) in favour of the three-spot model. Table 3 summarises our three-spot model MCMC calculations based on BTr, WIRE, SMEI, and photometry. The resulting maps are illustrated in Fig. 3. The tenth parameter listed in Table 3 is the rotation period with 90% credibility ranges. We note that the usual 68% interval for $P_{\text{rot}}$ is (for a Gaussian distribution) smaller by a factor of 1.645 (e.g. for BTr $\pm 0.0008 d$).

The BTr maps are based on 62735 BTr input data points obtained in 2016, and binned in 1678 time bins according to individual BRITE orbits, and are shown in the left column of Fig. 3.

The rather poor photometric quality of the SMEI data does not allow us to obtain a three-spot model because the MCMC algorithm simply does not converge: only two spots can be identified (Fig. 3 top in middle column). The surface map based on nearly 55 days of TESS (2019) photometry (Fig. 3 right column), confirms the map derived from BTr (2016) data. Overall, the TESS data quality is impressive and encourages looking for a fourth spot (cf. Table 3 and Fig. 3). The addition of another spot is rewarded by a substantial gain in goodness of fit. The residuals drop from $\pm 0.141$ mmag (3-spot model) to $\pm 0.125$ mmag.

A comparison of the maps in Fig. 3 middle column, on top (SMEI) and at the bottom (WIRE), seem to imply that the orientation of one map is flipped relative to the equator. This illustrates the difficulties one encounters when trying to convert a one-dimensional light curve into a two-dimensional surface map. While period and spot longitudes are comparatively well defined, estimating spot latitudes notoriously proves to be ill-posed unless the S/N is very high. Figure 4 illustrates the problem, showing two alternative WIRE maps that are only marginally inferior to the solution shown in Fig. 3 but with spot latitudes comparable to maps from BRITE and TESS data.

3.3.2. Blue maps

In the next step we reanalysed the much noisier blue data obtained with Bb*'. A reduced data quality affects the sophistication of models that can realistically be tested, and is expressed in the number of free parameters. For the noisy Bb' data we limited this number as much as possible by fixing all but one parameter, the brightness contrast (higher–lower: $\kappa$ is larger–smaller than 1), and being the same for all three spots. We found $\kappa = 1.05 \pm 0.04$, albeit with residuals as large as $\pm 5.9$ mmag (r.m.s.), when applying our red model ($\kappa = 1.25$) to the blue data. We recall that the blue photometry is about a factor of four noisier than the red photometry (Fig. 2 and Paper I).

When we increase the parameter space for a test by assigning each spot an individual $\kappa$, the blue spot 1 completely disappears, the fourth spot (cf. Table 4, and Fig. 3). The addition of another spot is rewarded by a substantial gain in goodness of fit. The residuals drop from $\pm 0.125$ mmag (3-spot model) to $\pm 0.108$ mmag.

Both of these models have comparable Bayesian evidence, and hence the solution converges to a two-spot model. Instead, when we formally allow a time lag $\Delta t$ in the three-spot model, the fit improves: $\kappa = 1.25 \pm 0.035$, and $\Delta t = 0.69 \pm 0.08$ d, with residuals of $\pm 5.1$ mmag (r.m.s.). The time lag of $\approx 17$ hours is equivalent to a phase shift of $56^\circ \pm 7^\circ$.

Both of these models have comparable Bayesian evidence, but whereas spots with different chemical compositions, and hence different $\kappa$ values, are known for CP stars, a colour-driven phase shift is not. In conclusion, the currently available blue pho-
Fig. 3. HJDs are transit times of the spot that contributes most to the light curve (spot 1). Left: BTr-based maps (top: mean values; bottom: modal values). Middle: SMEI mean map, which allows for only two spots (top) and WIRE (bottom), which indicates a three-spot map. Right: Comparison of TESS mean maps allowing for three spots (top) and four spots (bottom).

Fig. 4. Two maps derived from the same WIRE data. Both maps fit the light curve with an r.m.s. error only 0.002 mmag larger than the WIRE map presented in Fig. 3.

Photometry is insufficient for a reliable and convincing photometric surface mapping.

4. Pulsation of α Cir

In the following we use the same numbering of frequencies as was used by Bruntt et al. (2009) and in Paper I.

No pulsation is detected in the SMEI data with an amplitude exceeding 0.33 mmag, which corresponds to a S/N of 4.5. This is somewhat surprising given that the amplitude of the dominant pulsation frequency \( f_1 \) in the combined 2014 and 2016 BRITE-red data, with a similar effective wavelength, is equal to 0.57 mmag (Fig. 5). The amplitude in the WIRE data is equal to 0.65 mmag. However, the SMEI passband is very wide and corresponds to the sensitivity curve of a front-illuminated CCD, which ranges between 400 and 1000 nm (Table 1). Given different shapes, amplitudes, and phases of the rotational modulation

| Parameter | Value |
|-----------|-------|
| period (days) | 4.4799±0.0003 |
| residuals (r.m.s. in mmag) | ±0.125 |

**Notes.** Spots are ordered according to their impact on the light curve. 90% credibility limits are given to the values in the column ‘mean’. Spot longitudes are with respect to HJD 2458625.8065 (transit time of spot 1, see also Fig. 3, bottom of right column).
of α Cir at different wavelengths, it might well be that the pulsation signal has too small an amplitude in white-light SMEI data.

After subtracting the rotational modulation from the (un-binned) original time series, the Fourier amplitude spectrum shows a rich pattern of pulsation modes in the vicinity of \( f_1 \). However, a detailed analysis of this is beyond the scope of the present paper and we instead refer to a follow-up study that will present a detailed asteroseismic analysis (Kallinger et al., in preparation). Here, we use the oscillations in the TESS data primarily to guide and verify the analysis of the BRITE observations. For comparison we also show in Figs. 5 and 6 the Fourier spectrum of the combined 2006 WIRE data (∼170 d with a 114 d gap in between).

### 4.1. Primary pulsation frequency \( f_1 \)

After subtracting the rotational modulation via spline fits in the rotation-phase domain, the primary pulsation frequency \( (f_1) \) at about 210.99 d\(^{-1}\) is easily detected in all four BRITE data sets (Fig. 5), which we abbreviate as Bb*14 and Bb*16 (all blue filter data from 2014 and 2016) and BTr14 and BTr16 (all red filter data from 2014 and 2016). Combining the 2014 and 2016 data results in a total time base of ∼840 d with a coverage of about 35%. Merging the data improves the frequency resolution as well as the S/N, but as can be seen in Fig. 6 the ∼540-day gap between the two data sets causes strong aliasing with about ±0.0019d\(^{-1}\) and multiples.

To extract the oscillation parameters from the various light curves, we use a probabilistic approach presented by Kallinger & Weiss (2017). The automated Bayesian algorithm was originally developed to deal with multiple frequencies within the formal frequency resolution (Kallinger et al. 2017) but works with a mono-periodic signal (within one formal frequency resolution bin) as well. The method uses the nested sampling algorithm MultiNest (Feroz et al. 2009) to search for periodic signals in time series data and tests their statistical significance (i.e. not due to noise) by comparison with a constant signal. A solution is considered real if its probability \( p = z_{\text{signal}}/(z_{\text{signal}} + z_{\text{noise}}) > 0.9 \), where \( z \) is the global evidence delivered by MultiNest

Figure 5 shows highly significant peaks at about 210.9933d\(^{-1}\) in the combined BT and Bb* data, which compare well visually to those found in the WIRE and TESS data. We note that the BRITE-blue spectra amplitudes are divided by two in this figure for better comparison with the other data. The best-fit solutions are listed in Table 5. The

\[ \begin{align*}
\text{The global evidence is a normalised logarithmic probability describ-}
&\text{ing how well the model fits the data with respect to the uncertainties,}
&\text{parameter ranges, and the complexity of the fitted model.}
\end{align*} \]
uncertainties might appear unrealistically small, especially for the TESS data with a time base of less than 1/15 of the combined BRITE data. However, the frequency uncertainty is also influenced by the S/N of a frequency (e.g. [Kallinger et al. 2008]), which is about 18 times better in the TESS data than in the BRITE data.

While the frequencies extracted from the red and blue BRITE data agree exceptionally well with each other (within 0.00002 d$^{-1}$ or 30 µs, given a period of 409.491 7 s), there are small but significant differences to the frequencies found in the WIRE (0.0003 d$^{-1}$) and TESS (0.0018 d$^{-1}$) data. Such offsets might indicate a variable f₁, for example due to a companion or evolutionary effects, but aliasing in the BRITE data could also explain this (at least partly). Even though the strongest peaks in the BTr and Bb* data are nearly identical (Fig. 6), it might be that the physically relevant peak is one of the neighbouring alias peaks. If we consider the alias at +0.0019 d$^{-1}$ as the true frequency then it would almost perfectly resemble the TESS result. However, forcing our frequency analysis algorithm to fit this peak and comparing the resulting global evidence to the original one gives a probability of 0.97 against this scenario.

In conclusion, the average value for f₁, based on all BRITE data (years and colours) is 210.99328(2) d$^{-1}$. In addition, we find evidence that this frequency changes over time, but we leave a more detailed discussion to a follow-up study.

With the new BRITE data, we computed Fig. 7 which is analogous to Fig. 11 in Paper I, showing the amplitude and phase modulation of f₁ with the rotation phase. The phases given in this new figure differ from those in Paper I because different rotation periods have been used. The data are binned into ten rotation phases, where the starting epoch and the rotation period are the same as were used for phasing the rotational changes. The nearly sinusoidal shape of the amplitude modulation and the practically constant phase is consistent with an axisymmetric l = 1, m = 0 mode (Bruntt et al. 2009), with a rather small tilt between the pulsation and rotation axis (Bigot & Kurtz 2011).

### 4.2. Additional pulsation frequencies (f₂, f₇, and rotational side lobes of f₁)

After pre-whitening f₁ from the combined BRITE data we still find some signal left at frequencies where Bruntt et al. (2009) reported f₂ and f₇ in the WIRE data and which are unambiguously confirmed by TESS (Fig. 5). Even though the corresponding peaks barely exceed a S/N of three in the individual 2014 and 2016 BRITE data, they reach sufficient significance levels in the combined data (Table 5). Given the larger amplitude, f₁ reaches a probability of 0.99 (and therefore very strong evidence of not being due to noise) in both data sets. On the other hand, f₂ has a smaller amplitude and therefore also a lower probability of 0.91 (strong evidence) in the Bb* data and 0.77 (still substantial evidence) in the BTr data. We note that we would not consider the last peak as real on its own, but given its significance beyond doubt in all other time series we accept that it is...
Table 5. Detected frequencies in the various time series of α Cir.

|                | Frequency | Amplitude | Phase | p     |
|----------------|-----------|-----------|-------|-------|
|                | [d⁻¹]     | [parts per thousand] | [degrees] |       |
| f₁             | 210.99512(15) | 0.394(8) | -     | 1.0   |
| TESS           | 210.99361(5)  | 0.65(1)  | -     | 1.0   |
| WIRE           | 210.99329(2)  | 0.577(28) | 95(1) | 1.0   |
| BTr            | 210.99327(2)  | 1.716(81)| 84(1) | 1.0   |
| Bb*            |           |          |       |       |
| f₁ - f₁₀       | 211.2188(7)  | 0.039(3) | -     | 1.0   |
| f₁ + f₁₀       | 211.2171(1)  | 0.074(4) | -     | 1.0   |
| WIRE           | 211.2167(1)  | 0.334(86)| -     | 0.82  |
| BTb            | 208.3879(3)  | 0.089(3) | -     | 1.0   |
| Bb*            | 208.3866(2)  | 0.263(86)| -     | 0.31  |
| f₆             | 208.3866(1)  | 0.145(4) | -     | 1.0   |
| WIRE           | 208.3888(1)  | 0.098(30)| 23(7) | 0.77  |
| Bb*            | 208.3890(1)  | 0.318(82)| 11(5) | 0.91  |
| f₇             | 213.6005(3)  | 0.105(3) | -     | 1.0   |
| Bb*            | 213.6001(2)  | 0.375(83)| 76(7) | 0.99  |

Notes. The Bb* frequencies are derived from the combined BRITE-blue observations (Bb and Blb), obtained in 2014 and 2016. Phases are defined for HJD = 2457200 and are given in degrees. The parameter p gives the probability that a frequency is statistically significant compared to no signal (i.e. due to noise).

real in the BTr data. In a more Bayesian sense, one could use the TESS frequency (and its uncertainty) as a prior in the frequency analysis of the BTr data. This would significantly increase the peak’s probability of not being due to noise, but this approach considerably exceeds our present computational resources.

Even though f₀ and f₁ are easily detectable in the combined BRITE data, we find again an aliasing problem when determining their best-fit parameters. A closer look at Fig. 6 shows that while the strongest peak in the BTr spectrum is consistent with the position of f₁ in the TESS data, the dominant peak in the Bb* spectrum is shifted towards higher frequencies. However, the Bb* time series has several large gaps; the longest is about 12.9 d long, which causes aliases at about ±0.078 d⁻¹ (in addition to the ±0.0019 d⁻¹ aliases due to the large gap between the 2014 and 2016 data). If we now force our frequency analysis algorithm to fit the alias at −0.078 d⁻¹ instead of the largest-amplitude peak, we get very good agreement between the BTr and Bb* data. This strategy might appear arbitrary, but in fact it represents the Bayesian principle of prior information and is therefore justified.

We find a similar situation for f₀. While the largest-amplitude peak in the BTr spectrum agrees well with the position of the peak in the TESS data, the dominant peak in the Bb* spectrum appears to be a ±0.0019 d⁻¹ alias. A forced fit again aligns the frequencies we find in the blue and red BRITE data.

Finally, we also find some signal in the Bb* data at the expected rotational side lobes of f₁. The pulsation amplitude of an obliquely pulsating non-radial mode changes with rotation phase, as the aspect of the mode changes. This gives rise to frequency side lobes f₁ ± f₁₀ that describe the amplitude modulation (and phase variation, if present), which are clearly visible in the WIRE and TESS data. While the signal at f₁ is too weak to be distinguished from the noise (even though we know it has to be there), the peak at f₁⁺ is with p = 0.82 statistically significant. This allows for a seismic determination of the α Cir rotation period, which results in 4.4758 ± 0.0020 d, and is more accurate than those resulting from the WIRE and TESS data. Despite the greater noise of the BRITE data, the much longer time base permits a more accurate determination of Prot.

4.3. Mode identification and the large frequency separation

A frequently used observable in asteroseismic studies of high-overtone acoustic oscillations is the large frequency separation ∆νₙ₊₁, which is defined as the difference between modes of the same spherical degree and consecutive radian orders: ∆νₙ₊₁ = νₙ₊₁ − νₙ. The large separation becomes relevant for high radial orders, which are expected to follow the asymptotic relation (Tassoul 1980).

\[
\nu_{n,l} \approx \Delta \nu_{n,l}(n + 1/2 + \epsilon_0) - l(l + 1)D_0.
\]

Here \(\epsilon_0\) and \(D_0\) are parameters sensitive to the properties of the reflection layer near the stellar surface and the conditions in the stellar core, respectively. But more importantly, the large separation is related to the stellar acoustic diameter (i.e. inverse sound travel time across the stellar diameter). For an ideal adiabatic gas, this is proportional to the square root of the mean stellar
density. Consequently, $\Delta \nu$ provides a measure for the mean density of a star.

Even though a strong magnetic field has the tendency to distort spherical symmetry, the high-frequency oscillations observed in $\alpha$ Cir indicate that these modes are high-overtone acoustic oscillations for which one may expect to find a regular pattern corresponding to the asymptotic relation. As already noted by Bruntt et al. (2009), the three modes $f_1$, $f_6$, and $f_7$ are almost equidistant in frequency with an average separation of $\sim 2.606 \, d^{-1}$. One can naively expect that this value corresponds to the average frequency separation of $\alpha$ Cir. However, it is not compatible with other constraints on the stellar properties as we show in the following.

Bruntt et al. (2008) determined the angular diameter of $\alpha$ Cir to $\Theta_{LD} = 1.105 \pm 0.037 \, \text{mas}$, and used the revised Hipparcos parallax of van Leeuwen (2007) to estimate the star’s radius to be $1.97 \pm 0.07 \, R_\odot$. With an average frequency separation of $\sim 2.606 \, d^{-1}$ one can estimate the star’s mass according to $\Delta \nu \propto \sqrt{M/R^3}$. This is usually done by relating $\Delta \nu$ to the solar value of about $11.664 \, d^{-1}$ (e.g. Kallinger et al. 2010), which gives an unrealistically low mass of $0.38 \pm 0.04 \, M_\odot$. However, if we follow Bruntt et al. (2009) and set $\Delta \nu = 5.21 \, d^{-1}$ (i.e. $f_1 - f_6$), we obtain a plausible mass of $1.52 \pm 0.15 \, M_\odot$, which is also compatible with the estimate of $1.7 \pm 0.2 \, M_\odot$, based on the star’s position in the HR diagram Bruntt et al. (2008). We expect $\alpha$ Cir to have a slightly different mass than $1.5 \, M_\odot$ because the observed modes are magnetically distorted (e.g. Cunha 2006), which also affects the large separation compared to that of unperturbed modes for which the $\Delta \nu$ scaling is defined.

A consequence of $\Delta \nu$ being more likely equal to $5.21 \, d^{-1}$ than to $2.606 \, d^{-1}$ is that the three modes discussed above cannot be of the same spherical degree. This is also supported by Fig. 8 where we show the three modes in an Echelle diagram. While the modes almost perfectly line up vertically when folding their frequency with $2.6 \, d^{-1}$ (indicating the same spherical degree), this is not so when folding them with $5.2 \, d^{-1}$. In this case, $f_1$ is shifted by slightly more than one-half in the horizontal direction, while the other two modes have about the same horizontal offset. This is expected for a sequence of two consecutive $l = 0$ modes with an intermediate $l = 1$ mode, or vice versa (Eq. (1)). Assuming that the magnetic distortion is similar for all three modes, this also rules out the presence of a quadruple mode. Such a mode would need to be located closer to one mode than to the other (and not at their midpoint). Since Bruntt et al. (2009) argued that $f_1$ is very likely an axisymmetric dipole mode ($l = 1, m = 0$) based on simulated amplitude modulations for an oblique pulsator model, we conclude that $f_6$ and $f_7$ are very likely consecutive radial modes.

Even though a mode identification for roAp stars from multicolour photometry is notoriously difficult (e.g. Quirzl-Manosalva et al. 2018), further support is provided by the amplitude ratios and phase differences in the two BRITE passbands we determine for the three modes (Fig. 8). We expect that modes of the same spherical degree form clusters in a diagram like Figure 8. If we find modes that are clearly separated in the amplitude-ratio versus phase-difference plane, they can be assumed to have different spherical degrees. For $f_6$, the uncertainties are too large to make any conclusions, but $f_1$ and $f_7$ are separated by about $1.8 \, \sigma$ in their amplitude ratio, which corresponds to a probability of almost 0.9 that they are separated, and therefore have a different spherical degree.

5. Discussion

- **Data**: We analyse and discuss data obtained by BRITE-Constellation during 2014 and 2016, covering a total of 316 days and complement data volume and time base with observations archived from WIRE, SMEI, and the red and blue data. These data suffer from different noise levels and instrumental effects, which need to be considered and corrected. Corrections of trends and averaging procedures need to be optimised for a discussion of long-term (rotation, spots) and short-term (pulsation) effects. An attractive aspect of the data is the availability of data in two passbands, some of them taken nearly simultaneously. This allowed us to study in depth the pros and cons of Bayesian based photometric surface imaging.

Not surprisingly, the photometric quality of the data seriously limits the number of detectable spots and their location, as is discussed in Section 3. Spot locations are well constrained by the data; instead, latitudes are, photometrically, a notoriously ill-defined parameter. In the case of Doppler imaging, the situation is much better due to the availability of additional information.

- **Rotation**: The determination of $P_{\text{rot}}$ for $\alpha$ Cir dates back to 1991 (Mathys 1991). Meanwhile, many photometric, spectroscopic, and polarimetric observation were performed and we determined the formally most accurate $P_{\text{rot}}$ from combined WIRE and SMEI spot transit times to be $4.47930 \pm 0.00002 \, d$ (68% interval).

The different shape of the rotation-phase plot extracted from the red and blue data indicates a different chemical composition of at least one of the spots, which, however, can only be tested by high-resolution spectroscopy.

- **Spots**: Two spots were identified in Paper I in the red data from 2014, but with better data obtained with BTR in 2016 we have clear evidence of three spots. Despite three more degrees of freedom, the gain in evidence from a two-spot model to a three-spot model
spot model is substantial, although the residuals to the model light curves decreased only from 1.389 to 1.386 mmag. In comparison, the much noisier data from SMEI (up to a factor of 4) allow only two spots to be considered.

Bayesian photometric imaging routines result in numerous solutions. As an example, Fig. 3 shows two solutions to the BTr (2016) data: the barycentric solution (top, mean values) and best-fit solution (bottom, modal values). They do not coincide satisfactorily, which hints at a serious non-Gaussianity of the posterior, as is illustrated in Fig. 1. The most striking difference between the Bayesian photometric imaging of the BTr (2016) photometry and the WIRE data set (see Fig. 5) concerns spot latitudes. The marginal distribution of the spot 1 latitude for the BTr data seems to exclude a northern location, contrary to what is determined for the WIRE data. The ‘north pole’ is defined as the pole that is visible from Earth.

Here a cautionary note should be heeded: One should keep in mind that the often surprisingly narrow marginal distribution is due to a model’s rigidity. It measures its ‘elbow room’ and nothing else. If one could relax the rigidity of the model, for example by allowing non-circular spots and/or variable contrasts, the marginal distribution would spread out because of the increased freedom. But any additional degrees of freedom come at a price: they would reduce the evidence of a three-spot model. There is a trade-off between precision of a fit and its credibility. We adhere here to Fermi’s rule: it is better to be approximately right than precisely wrong.

- **Pulsation**: No pulsation is detected in the SMEI data with an amplitude exceeding the detection threshold of 0.33 mmag, corresponding to a S/N = 4.5. The dominant pulsation frequency $f_1$ derived from BRITE-red data agrees exceptionally well with that derived from -blue data, but there are significant differences to the pulsation frequencies derived from WIRE and -which will be subject to a follow-up paper with a detailed asteroseismic analysis (Kallinger et al., in preparation). We can improve the pulsation frequency by combining the times of maximum from BRITE-red and WIRE data, resulting in $f_1 = 210.993264(5) \text{ d}^{-1}$, which is the most accurately determined $f_1$ to date for any roAp star. The three hitherto well-established frequencies very likely come from two consecutive radial $l = 0$ modes ($f_0$ and $f_1$), with an intermediate $l = 1$ mode ($f_2$).

6. Conclusion

At least three surface spots can be identified for α Cir, which confirms the conclusion of Paper I that a two-spot model is too simple. The high-quality "data even suggest a fourth spot. On the other hand, the low-quality blue data barely indicate a spot at rotation phase 0.6 (Fig. 3).

According to our experience the best-fit set of parameters, indicated by a minimum $\chi^2$, differs significantly from the set of mean values (inferred from the marginal distributions of the parameters), which hints at a noticeable skewness of the probability distribution in the ten-dimensional configuration space considered. Spot latitudes are, as expected, less well determined than longitudes. To our knowledge, this is the first time that a Bayesian-based evidence of models differing in the number of spots has been quantitatively determined.

Concerning the main pulsation frequency of α Cir, we were able to improve the accuracy to 60 Hz (0.01 msec for a 6.825 min period), assuming a stable frequency.

The photometric data obtained to date for α Cir clearly illustrate the need for high-precision data on the one hand and long data sets on the other. In general, both qualities are needed for convincing astrophysical analyses, and this should be considered when deciding about investments in ground-based and in space-based instrumentation.

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