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Sub- $m s^{-1}$ upper limits from a deep HARPS-N radial-velocity search for planets orbiting HD 166620 and HD 144579*

A. Anna John ^{1,2} † A. Collier Cameron ^{1,2} J. P. Faria,³ A. Mortier ⁴ T. G. Wilson ^{1,2}
 L. Malavolta,^{11,12} L.A. Buchhave,⁵ X. Dumusque,⁸ M. López-Morales ⁹ R.D. Haywood ^{7,13}
 K. Rice ^{6,14} A. Sozzetti ¹⁰ D. W. Latham,⁹ S. Udry,¹⁵ F. Pepe,²⁰ M. Pinamonti,¹⁰ A. Vanderburg,²¹
 A. Ghedina,¹⁸ R. Cosentino,¹⁸ M. Stalport,^{19,20} B. A. Nicholson,^{16,17} A. Fiorenzano ¹⁸ E. Poretti ¹⁸

¹*SUPA, School of Physics & Astronomy, University of St Andrews, North Haugh, St Andrews, KY169SS, UK*

²*Centre for Exoplanet Science, University of St Andrews, North Haugh, St Andrews, KY169SS, UK*

³*Instituto de Astrofísica e Ciências do Espaço, Universidade do Porto, CAUP, Rua das Estrelas, 4150-762 Porto, Portugal*

⁴*School of Physics & Astronomy, University of Birmingham, Edgbaston, Birmingham B15 2TT, UK*

⁵*DTU Space, National Space Institute, Technical University of Denmark, Elektrovej 328, DK-2800 Kgs. Lyngby, Denmark*

⁶*SUPA, Institute for Astronomy, University of Edinburgh, Blackford Hill, Edinburgh, EH9 3HJ, UK*

⁷*Astrophysics Group, University of Exeter, Exeter EX4 2QL, UK*

⁸*Observatoire Astronomique de l'Université de Genève, Chemin Pegasi 51, 1290 Versoix, Switzerland*

⁹*Center for Astrophysics | Harvard & Smithsonian, 60 Garden Street, Cambridge, MA 02138, USA*

¹⁰*INAF - Osservatorio Astrofisico di Torino, Via Osservatorio 20, I-10025 Pino Torinese, Italy*

¹¹*Dipartimento di Fisica e Astronomia "Galileo Galilei" - Università di Padova, Vicolo dell'Osservatorio 3, 35122, Padova, Italy;*

¹²*INAF - Osservatorio Astronomico di Padova, Vicolo dell'Osservatorio 5, 35122, Padova, Italy*

¹³*STFC Ernest Rutherford Fellow*

¹⁴*Centre for Exoplanet Science, University of Edinburgh, Edinburgh, EH9 3FD, UK*

¹⁵*Observatoire de Genève, Université de Genève, 51 chemin des Maillettes, 1290 Sauverny, Switzerland*

¹⁶*University of Southern Queensland, Centre for Astrophysics, Toowoomba, Australia, 4350*

¹⁷*Sub-department of Astrophysics, University of Oxford, Keble Rd, Oxford, United Kingdom, OX13RH*

¹⁸*Fundación Galileo Galilei – INAF, Rambla José Ana Fernandez Pérez 7, 38712 – Breña Baja, Spain*

¹⁹*Space sciences, Technologies and Astrophysics Research (STAR) Institute, Université de Liège, Allée du 6 Août 19C, 4000 Liège, Belgium*

²⁰*Département d'Astronomie, Université de Genève, Chemin Pegasi 51b, CH-1290 Versoix, Suisse*

²¹*Department of Physics and Kavli Institute for Astrophysics and Space Research, Massachusetts Institute of Technology, 77 Massachusetts Avenue, Cambridge, MA 02139, USA*

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ABSTRACT

Minimising the impact of stellar variability in Radial Velocity (RV) measurements is a critical challenge in achieving the 10 $cm s^{-1}$ precision needed to hunt for Earth twins. Since 2012, a dedicated programme has been underway with HARPS-N, to conduct a blind RV Rocky Planets Search (RPS) around bright stars in the Northern Hemisphere. Here we describe the results of a comprehensive search for planetary systems in two RPS targets, HD 166620 and HD 144579. Using wavelength-domain line-profile decorrelation vectors to mitigate the stellar activity and performing a deep search for planetary reflex motions using a trans-dimensional nested sampler, we found no significant planetary signals in the data sets of either of the stars. We validated the results via data-splitting and injection recovery tests. Additionally, we obtained the 95th percentile detection limits on the HARPS-N RVs. We found that the likelihood of finding a low-mass planet increases noticeably across a wide period range when the inherent stellar variability is corrected for using SCALPELS **U**-vectors. We are able to detect planet signals with $M \sin i \leq 1 M_{\oplus}$ for orbital periods shorter than 10 days. We demonstrate that with our decorrelation technique, we are able to detect signals as low as $54 cm s^{-1}$, which brings us closer to the calibration limit of $50 cm s^{-1}$ demonstrated by HARPS-N. Therefore, we show that we can push down towards the RV precision required to find Earth analogues using high-precision radial velocity data with novel data-analysis techniques.

Key words: planets and satellites: detection, techniques: radial velocities, stars: activity, line: profiles, individual: Rocky Planet Search targets HD 166620 and HD 144579.

* Results presented in this paper have been obtained within the HARPS-N Collaboration.

† E-mail: aaj1@st-andrews.ac.uk

1 INTRODUCTION

A Radial Velocity (RV) precision of 0.1 m s^{-1} is essential for the detection of Earth-sized low-mass exoplanets orbiting in the habitable zones of Sun-like stars. Each subsequent advancement in instrumental stability since the discovery of the first exoplanet around a Sun-like main sequence star in 1995, has been reflected in a reduction in the detection limit of exoplanets down to 1 m s^{-1} (green shaded region in Figure 1). Thanks to improvements in precision, it is now possible to use the new generation of spectrometers to observe the reflex motions of stars hosting such low-mass planets. However, the red-shaded region in Figure 1 shows that despite these further instrumental advancements, the declining trend in the detectable RV semi-amplitude has hit a " 1 m s^{-1} saturation limit", at an amplitude comparable to RV variations produced by the intrinsic variability of the host star.

Stellar variability is an umbrella term for a variety of astrophysical processes that have an impact on the RV of stars on timescales ranging from minutes to days. While stellar p -modes impose RV variability of up to 1 m s^{-1} on a minutes scale in solar-type stars (e.g., Chaplin et al. 2019; Bouchy, Pepe, & Queloz 2001), granulation introduces variability on timescales of hours to days (e.g., Gray 2009; Meunier et al. 2015). Astronomers combat these parasitic signals by using longer exposure times and averaging out (for p -modes), and observing the stellar target multiple times throughout the night at intervals of several hours (for granulation) (Dumusque et al. 2011). On magnetically active stars, starspots and faculae/plage can cause RV shifts of the order of $1\text{--}100 \text{ m s}^{-1}$, while the suppression of convection causes shifts of about $10\text{--}20 \text{ m s}^{-1}$ (Meunier, Desort, & Lagrange 2010). Magnetic activity poses the biggest obstacle to the identification and characterization of small exoplanets, as it can occur on similar timescales and can completely obscure or mimic the RV signals of real exoplanets.

Recent studies have had some success modelling stellar variability by employing photometry to independently estimate the impact of stellar activity on the RVs (Aigrain, Pont, & Zucker 2012). Other approaches include using time domain techniques like Gaussian Process (GP) regression to model the correlated noise induced by stellar activity (e.g., Haywood et al. 2014; Rajpaul et al. 2015; Delisle, Hara, & Ségransan 2020) and using multidimensional GP to use spectroscopic activity-indicators in conjunction with RVs to constrain the activity-induced signal in the RV time-series (e.g., Rajpaul et al. 2015; Barragán et al. 2022). Using neural networks to distinguish activity signals from true center-of-mass RV shifts of dynamical origin (de Beurs et al. 2021) and de-trending the RVs for line shape variations using the SCALPELS basis vectors (Collier Cameron et al. 2021) or spectral SHELL vectors (Cretignier, Dumusque, & Pepe 2022) are a few wavelength domain techniques. Doppler imaging is also used to model both activity-induced variations and planet-induced shifts simultaneously in the spectral line profiles (Klein et al. 2022). A self-consistent comparison of these different methods (and more) is presented in Zhao et al. (2022).

In this paper, we present a deep search for planetary reflex-motion in the HARPS-N data of bright K & G dwarfs HD 166620 and HD 144579, obtained as part of the Rocky Planet Search programme. We use TWEAKS (Time and Wavelength-domain stEllar Activity mitigation using KIMA and SCALPELS): an RV analysis pipeline designed for attaining sub- m s^{-1} detection thresholds at long orbital periods, by combining wavelength-domain and time-domain stellar activity mitigation (Anna John, Collier Cameron, & Wilson 2022). This approach takes the posterior probability distributions for the orbital parameters of a system determined along with a complete Keplerian solver and a Gaussian Process (on/off) using the nested sampling al-

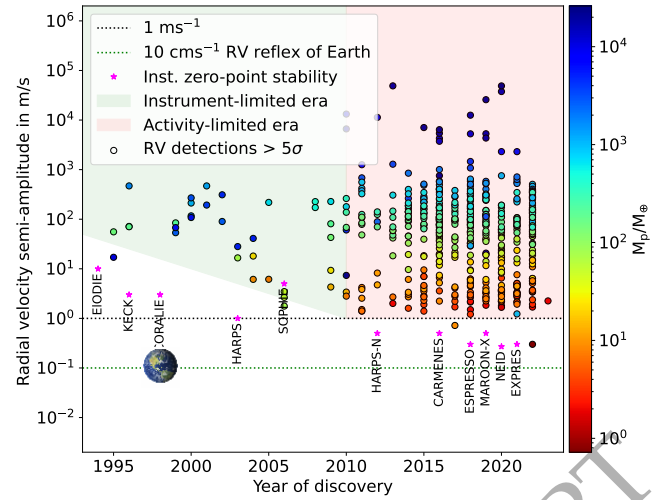


Figure 1. The radial velocity semi-amplitudes of planets obtained from the NASA Exoplanet Archive are shown as a function of the discovery year. The green-shaded region shows an era in which the detection threshold improved almost exponentially in response to the improvement in wavelength calibration. However, this downward trend stops and flattens out in the early 2000s, despite the significant improvement in the instrument precision achieved by the new generation spectrographs.

gorithm, KIMA (Faria et al. 2018). We then use SCALPELS basis vectors for spectral line-shape decorrelation (Collier Cameron et al. 2021), following the result from Anna John, Collier Cameron, & Wilson (2022) which shows that de-trending the RVs for line shape variations using the SCALPELS basis vectors yields a significantly better RV model over a model that does not account for these stellar activity signatures. There are also other studies that use SCALPELS and TWEAKS for improved detection of exoplanets (Wilson et al. (2022), Stalport 2023 submitted, Palethorpe 2023 in prep., Dalal 2023 in prep.).

2 TARGETS & OBSERVATIONS

We applied TWEAKS to two targets: HD 166620 and HD 144579. HD 166620 and HD 144579 have been observed for more than a decade with HARPS-N (Cosentino et al. 2012), as part of the HARPS-N collaboration Rocky Planet Search (RPS) initiative. The RPS programme was initiated in 2012 to search for small planets orbiting bright, nearby stars (mainly K and early M dwarfs) when the Kepler field was unobservable. The major objective of the RPS program is to conduct a systematic search for low-mass planets around nearby quiet stars that are visible from the northern hemisphere via extensive monitoring of the RV, with extremely high precision (Moutalebi et al. 2015). Targets for this programme are monitored 2-3 times per night (to mitigate granulation), with exposure times of 15 minutes, which is the cumulative open shutter time during multiple sub-exposures taken to prevent saturation of the brightest spectral orders. This helps to average out the p -mode oscillations (Chaplin et al. 2019), and thereby reduce the impact of stellar variability with brief typical timescales.

A total of 1025 observations of HD 166620 were obtained between 2013 March and 2023 March, and 888 observations of HD 144579 were taken between 2013 February and 2022 March. For the mitigation of p -mode oscillations and granulation, we use the daily

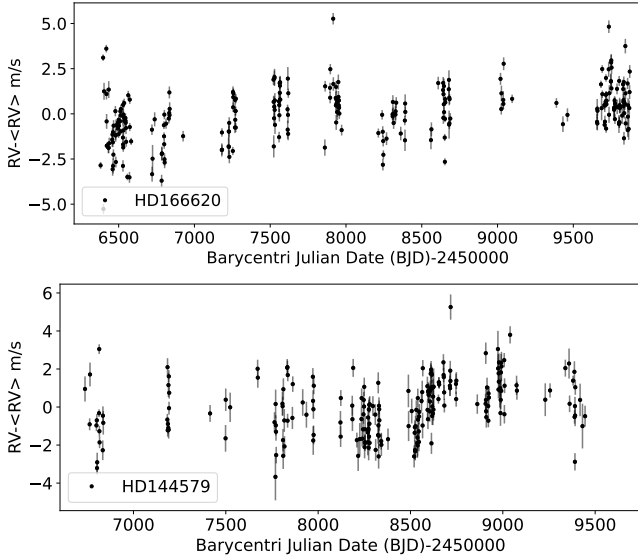


Figure 2. Nightly binned radial velocity observations (median-subtracted) for HD 166620 and HD 144579 measured from HARPS-N spectra are shown in the top and bottom panels, respectively.

binned observations in this study, which reduces the total number of RV observations for HD 166620 and HD 144579 to 293 and 256, respectively (Figure 2). After binning, the radial velocity RMS for HD 166620 reduced from 1.8 to 1.45 m s⁻¹ and a reduction from 3.5 to 1.29 m s⁻¹ was observed for HD 144579.

The HARPS-N Data Reduction Software (DRS), version 2.3.5 (Dumusque et al. 2021), was used to reduce the spectra used in this research. The DRS uses a mask (K2 & G9 spectral types were used for HD 166620 and HD 144579, respectively) composed of a multitude of lines that span the entire HARPS-N spectral range to calculate the Cross-Correlation Function (CCF), which is a representation of the average shape of the absorption profiles across the whole spectrum. Fitting a Gaussian to the CCFs then determines the stellar radial velocity, Full-width at Half Maximum (FWHM) and CCF Area (product of FWHM and the central line depth). In this analysis, we use the CCFs as they are obtained from the DRS. The DRS pipeline also provides supplementary information such as the spectral bisector span (BIS), H α , Na- Index, and Mount Wilson S-index for each observation. These additional data products are widely used as stellar activity indicators.

HD 166620 is a K2V type star with $V_{\text{mag}} = 6.38$ (Motalebi et al. 2015). The Mount Wilson program (1966–2001) (Duncan et al. 1991) studied the stellar chromospheric activity of HD 166620, leaving us with an extensive record of the stellar chromospheric activity (Baliunas et al. 1995). Baum et al. (2022) updated the stellar parameters for HD 166620 and identified this star as a possible Maunder minimum candidate based on the five decades of CaII H&K measurements from the Mount Wilson Survey, and continued observations at Keck as part of the California Planet Search programme (Wright 2004; Isaacson & Fischer 2010). Recently, Luhn et al. (2022) presented additional Mount Wilson data that definitively traced the transition from cyclic activity to a prolonged phase of flat activity.

Baum et al. (2022) and Luhn et al. (2022) concluded HD 166620 to be the first unambiguous Maunder-Minimum analogue, as shown by its S-index activity time series, as it changed from a cycling to a flat activity state. We reproduce Luhn et al. (2022)’s data along

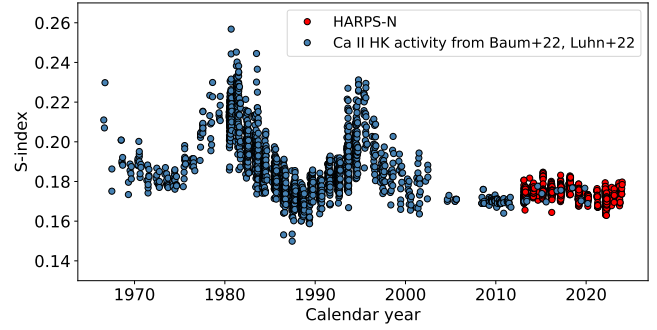


Figure 3. The Ca II H&K S-index data of HD 166620 from Mount Wilson and Keck-Hires, displaying the transition into the Maunder Minimum (e.g., Baum et al. 2022; Luhn et al. 2022). The 10 years of S-index data obtained from HARPS-N, shown by the overlapping red points reinforces the star’s present Maunder minimum phase. We adjusted for the offset in the HARPS-N data with a constant shift to have a consistent median S-index value with the Mount Wilson data set. Altogether we present 57 years of stellar chromospheric activity record for HD 166620 here.

Table 1. Stellar parameters obtained for HD 166620 and HD 144579 from three independent methods. Our final adopted parameters, listed in Table 2, are the weighted average of the results from these three methods.

Parameter	HD 166620	HD 144579	Method
T_{eff} (K)	5005 ± 116	5323 ± 70	ARES+MOOG
	4966 ± 70	5248 ± 70	CCFPams
	4997 ± 50	5316 ± 50	SPC
$\log(g)$ (cgs)	4.82 ± 0.22	4.84 ± 0.12	ARES+MOOG
	4.70 ± 0.20	4.56 ± 0.20	CCFPams
	4.43 ± 0.10	4.60 ± 0.10	SPC
[Fe/H]	-0.21 ± 0.06	-0.63 ± 0.05	ARES+MOOG
	-0.25 ± 0.05	-0.70 ± 0.05	CCFPams
[m/H]	-0.19 ± 0.08	-0.62 ± 0.08	SPC

with the 10 years of S-index data from HARPS-N here in Figure 3. This ‘flattened’ state is explained by Luhn et al. (2022) as a less active phase of the star with fewer residual active areas to account for the even lower variability. This span of low activity includes the period of time from 2012 to the present, during which the radial velocity of HD 166620 has been monitored intensively for planetary reflex-motion as part of the HARPS-N Rocky Planet Search (RPS) programme.

The second target of this study is HD 144579. In contrast to HD 166620, no in-depth literature is presently available about this extremely bright G8V star with $V_{\text{mag}} = 6.65$. The stellar parameters obtained as part of this study together with the Gaia Early Data Release (Gaia Collaboration et al. 2021) for both the stars are given in Section 2.1 and Table 2.

2.1 Stellar characteristics of HD 166620 and HD 144579

Galactic velocities for both stars were derived using the Gaia DR3 (Gaia Collaboration et al. 2021) data and in particular the trigonometric parallax, the radial velocity, the positions and the proper motions. Following Johnson & Soderblom (1987) we calculated the galactic velocities U , V , and W , as reported in Table 2. These are expressed in the directions of the Galactic center, Galactic rotation, and north Galactic pole, respectively, and we did not subtract the solar motion

Table 2. Stellar parameters obtained for HD 166620 and HD 144579 from the literature and from the present study. All methods obtaining the parameters are described in Section 2.1

Parameter	HD 166620	HD 144579	Reference
RA (J2000)	18:09:37.4162	16:04:56.7936	Gaia Collaboration et al. (2021)
DEC (J2000)	+38 27 27.9980	+39:09:23.4346	Gaia Collaboration et al. (2021)
μ_α (mas/yr)	-316.454 ± 0.018	-570.872 ± 0.016	Gaia Collaboration et al. (2021)
μ_δ (mas/yr)	-468.348 ± 0.020	52.633 ± 0.017	Gaia Collaboration et al. (2021)
G	6.130 ± 0.003	6.461 ± 0.003	Gaia Collaboration et al. (2021)
B	7.28 ± 0.02	7.388 ± 0.065	Wenger et al. (2000)
V	6.39 ± 0.02	6.655 ± 0.014	Wenger et al. (2000)
J	4.952 ± 0.0278	5.182 ± 0.02	Cutri et al. (2003)
H	4.458 ± 0.192	4.824 ± 0.017	Cutri et al. (2003)
K	4.232 ± 0.021	4.755 ± 0.016	Cutri et al. (2003)
$W1$	4.236 ± 0.279	4.733 ± 0.178	Cutri et al. (2021)
$W2$	3.996 ± 0.13	4.601 ± 0.083	Cutri et al. (2021)
$W3$	4.235 ± 0.014	4.777 ± 0.015	Cutri et al. (2021)
Spectral type	K2V	G8V	This work
Parallax (mas)	90.12 ± 0.02	69.64 ± 0.014	Gaia Collaboration et al. (2021)
Radial Velocity (km s ⁻¹)	-19.51 ± 0.12	-59.44 ± 0.12	Gaia Collaboration et al. (2021)
U (km s ⁻¹)	16.80 ± 0.05	-35.88 ± 0.04	This work
V (km s ⁻¹)	-31.34 ± 0.10	-58.51 ± 0.07	This work
W (km s ⁻¹)	0.25 ± 0.05	-18.57 ± 0.09	This work
T_{eff} (K)	4989 ± 48	5296 ± 37	This work ^a
$\log(g)_{\text{spec}}$ (cgs)	4.65 ± 0.10	4.67 ± 0.08	This work ^a
[Fe/H] (dex)	-0.21 ± 0.04	-0.65 ± 0.04	This work ^a
[α /Fe]	0.20	0.26	This work
$v \sin i$ (km s ⁻¹)	< 2.0	< 2.0	This work
ξ_t (km s ⁻¹)	0.45 ± 0.31	0.62 ± 0.12	This work
Mass (M_\odot)	$0.76^{+0.032}_{-0.019}$	$0.73^{+0.022}_{-0.013}$	This work
Radius (R_\odot)	$0.77^{+0.007}_{-0.006}$	$0.76^{+0.005}_{-0.004}$	This work
$\log g$ (cgs)	$4.55^{+0.02}_{-0.01}$	4.53 ± 0.01	This work
Density (ρ_\odot)	$1.67^{+0.093}_{-0.059}$	$1.63^{+0.057}_{-0.038}$	This work
Luminosity (L_\odot)	$0.36^{+0.02}_{-0.01}$	0.47 ± 0.01	This work
Distance (pc)	11.09 ± 0.002	14.36 ± 0.002	This work
Age (Gyr)	$10.09^{+2.73}_{-3.76}$	$11.90^{+1.57}_{-2.50}$	This work
$\log(R' \text{ HK})$ (median)	-5.143	-4.976	This work
P_{rot}	42.4 d	–	Vaughan et al. (1981)
P_{cyc}	17 years	–	Baum et al. (2022)

from our calculations. Using this kinematical information, we can estimate which population in the Galaxy our stars belong to. Using the descriptions in Reddy, Lambert, & Allende Prieto (2006), we find that there is 98.5% and 90.8% probability that HD 166620 and HD 1445797, respectively, belong to the thin disc.

We utilised our HARPS-N spectra (details below) to estimate the stellar atmospheric parameters: effective temperature (T_{eff}), metallicity [Fe/H], surface gravity, microturbulent velocity and projected rotational velocity. Following Mortier et al. (2020), three independent methods were used for obtaining these parameters: (1)

ARES+MOOG¹, using a stacked spectrum for each star, where the stitched spectra were simply added together after putting them in the lab-frame. Parameters were estimated using a curve-of-growth method based on neutral and ionised iron lines (Sousa 2014). (2) CCFpams², using the Cross Correlation Functions and applied an empirical relation, found in Malavolta et al. (2017). (3) Finally, spectral synthesis was applied on all individual spectra using the Stellar Parameter Classification tool (SPC - Buchhave et al. 2012).

Surface gravities are notoriously hard to measure accurately from spectra. The values resulting from ARES+MOOG and CCFpams were corrected following Mortier et al. (2014) while the value for SPC was additionally constrained using YY isochrones (Yi et al. 2001). Precision errors for effective temperature, surface gravity and metallicity from ARES+MOOG and CCFpams were inflated for accuracy by 60 K, 0.1 dex and 0.04 dex respectively following Sousa et al. (2011). Errors for the values of SPC are also accounted for accuracy (Buchhave et al. 2012). We note that ARES+MOOG is the only method to derive microturbulent velocity (ξ_t) while SPC is the only method that derived the projected rotational velocity ($v \sin i$).

The stacked spectra were additionally used to derive individual chemical abundances for magnesium, silicon and titanium. To calculate these abundances, ARES+MOOG was used with more details provided in Mortier et al. (2013). These three abundances, in combination with the iron abundance, were then used to calculate the alpha-over-iron abundance, $[\alpha/\text{Fe}]$. We find that HD 166620 has an alpha element overabundance of 0.20 and HD 144579 of 0.26, calculated with respect to iron as compared to the Sun. That makes these stars as alpha-enhanced as the planet hosts K2-111 (Mortier et al. 2020) or TOI-561 (Lacedelli et al. 2021) whose small planets have densities that are significantly below the densities of small planets around non-alpha-enhanced stars.

Stellar mass, radius, density, luminosity, distance, and age were further obtained from an isochrones and evolutionary tracks analysis, as described in Mortier et al. (2020). We used the three individual values for effective temperature and metallicity, the Gaia DR3 parallax, apparent magnitude in 8 photometric filters (listed in Table 2), and two different sets of isochrones (Dartmouth and MIST). For each parameter, six sets of posterior distributions were obtained using the code `isochrones` (Morton 2015) and the nested sampling algorithm `MultiNest` (Feroz et al. 2019). All six scenarios yielded consistent results. The final adopted parameters, as listed in Table 2, were calculated from the median and 16th and 84th percentile of the merged posterior distributions. A more precise value for surface gravity was then calculated using the stellar mass and radius.

3 ANALYSES AND RESULTS

3.1 RV analysis pipeline

We employed `TWEAKS` as an RV analysis pipeline to search for planetary signals below the 1 m s^{-1} RV barrier imposed by stellar activity (Figure 1). This pipeline uses the `SCALPELS` technique to separate the stellar variability component in the RVs, driven by spectral line-shape changes (Collier Cameron et al. 2021). The temporal variability of the CCF contains both spectral line ‘shape’ changes and Doppler ‘shifts’. `SCALPELS` uses the translational invariance property of the autocorrelation function (ACF; Adler & Konheim (1962)) of the CCF to

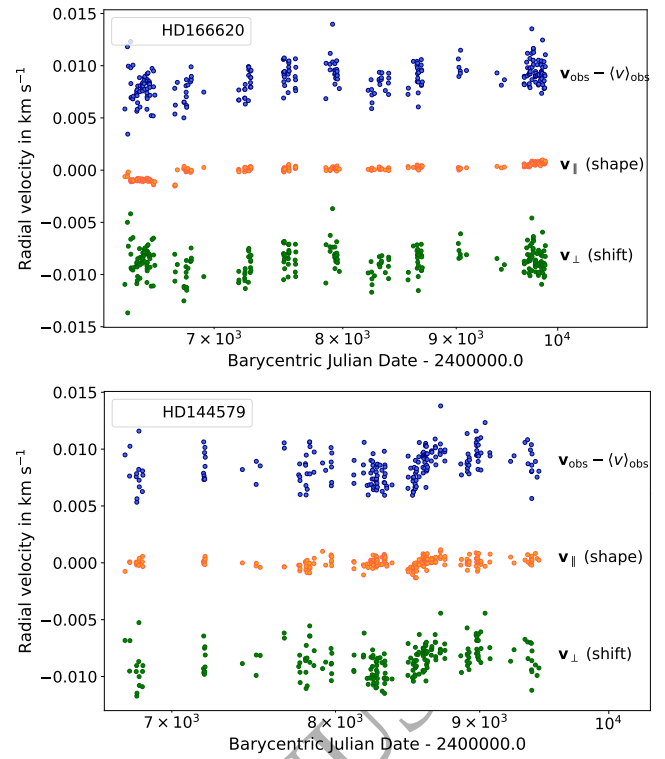


Figure 4. In both panels, the blue scatterplot at the top is the barycentric RV with its own mean subtracted. The orange time series in the middle represents the shape-driven component obtained from the `SCALPELS` projection, while the green one shows the ‘cleaned’ shift-driven velocities, obtained when the shape-driven variations are subtracted from the observed RVs. Please note that an offset is introduced to each RV component for better illustration. The jump in the shape time series of HD166620, at around BJD= 2456737 corresponds to the focus intervention that occurred in the HARPS-N instrument.

isolate the effects of shape changes in the CCF from shifts of dynamical origin. `SCALPELS` constructs an orthogonal basis containing the coefficients of the first few principal components of shape-induced CCF variations, then projects the raw RVs onto this basis to obtain a time series of shape-induced RV variations. These are then subtracted from the original RVs to leave shift-only RV variations, where the planet signals are sought (Anna John, Collier Cameron, & Wilson 2022).

The perturbations resulting from stellar activity (shape component, shown as the orange time-series in Figure 4) are isolated by projecting the RVs onto the time-domain subspace spanned by the amplitude coefficients (a.k.a \mathbf{U} -vectors) of the ACF’s principal components (Collier Cameron et al. 2021). However, when projected onto the orthogonal complement of the time-domain (\mathbf{U}) subspace, the dynamical shifts (shift component, shown as the green timeseries in Figure 4) due to the planets are preserved.

We found that the first three leading principal components are the strong contributors to the shape component out of the first five (see Figure D2 & D1). This optimal ranking was done using leave-one-out cross-validation (LOOCV). Collier Cameron et al. (2021) found this method to be efficient in identifying the number of leading columns of \mathbf{U} that contain significant profile information. The raw implementation of `SCALPELS` returns not just the basis vectors, but also the RV response. However, the shape time series in Figure 4 are from the `SCALPELS`-only implementation, not the coefficients of the amplitude obtained from the simultaneous fitting of any keplerian.

¹ ARESv2: <http://www.astro.up.pt/~sousasag/ares/>; MOOG 2017: <http://www.as.utexas.edu/~chris/moog.html>

² <https://github.com/LucaMalavolta/CCFpams>

Table 3. Prior distributions used in the radial velocity model in the KIMA analysis

Parameter	Notation	Unit	HD 166620	HD 144579	Distribution
Number of Keplerians to fit	N_p		$\mathcal{U}(0, 5)$	$\mathcal{U}(0, 5)$	Uniform
Orbital period	P	days	$\mathcal{U}(1.1, 3492)$	$\mathcal{U}(1.1, 2700)$	Uniform
RV Semi-amplitude	K	m s^{-1}	$\mathcal{MLU}(0.2, 15)$	$\mathcal{MLU}(0.2, 4)$	Modified Log-Uniform
Eccentricity	e		$\mathcal{K}(0.867, 3.03)$	$\mathcal{K}(0.867, 3.03)$	Kumaraswamy
Orbital phase	ϕ		$\mathcal{U}(0, 2\pi)$	$\mathcal{U}(0, 2\pi)$	Uniform
Longitude of line of sight	ω		$\mathcal{U}(0, 2\pi)$	$\mathcal{U}(0, 2\pi)$	Uniform
Decorrelation vectors	β	m s^{-1}	$\mathcal{G}(0, 10)$	$\mathcal{G}(0, 10)$	Gaussian
Systemic velocity	γ	m s^{-1}	$\mathcal{U}(-2.89, 5.37)$	$\mathcal{U}(-4.84, 4.96)$	Uniform
White noise	σ_{jit}	m s^{-1}	$\mathcal{LU}(0.01, 10 \times \text{rms})$	$\mathcal{LU}(0.01, 10 \times \text{rms})$	Log Uniform

The basis vectors (**U**-vectors) representing the shape components identified by SCALPELS are then used for stellar activity decorrelation using the KIMA nested sampling package (Faria et al. 2018). KIMA employs Diffusive Nested Sampling (DNS Brewer (2014)) to sample the posterior distributions for each of the orbital parameters by modelling the RV data with a sum of up to N_p Keplerian functions. This gives KIMA the estimates for the Bayesian evidence (\mathcal{Z}) of the data supporting each model, enabling the model comparison (Brewer, Pártay, & Csányi 2010). The ratio of Bayesian evidences between models with different N_p can then be compared. There is a ‘known’ object setting provided by KIMA, which allows one set of priors to be applied to the orbital parameters of a known planet while a different set of priors is used to explore the existence and parameters of putative planets (Standing et al. 2022). When a candidate Keplerian signal is proposed in the model, KIMA conducts fits for both the ‘known’ planet and the new signal. This option is useful if there is a transiting planet with precisely determined transit parameters or another ‘known’ Keplerian signal in the system that we can confidently fit out. Anna John, Collier Cameron, & Wilson (2022), presented a pilot demonstration of TWEAKS in action. They used a GP along with SCALPELS + KIMA to update the transiting planet’s mass and establish the existence of a third planet in the CoRoT-7 system.

KIMA posteriors are also used to estimate the detection limits as in Standing et al. (e.g., 2022, 2023) and Sairam & Triard (2022) for the entire orbital period range spanned by the data. With KIMA, we test our sensitivity to detecting low-mass planets and use it to generate a reliable limit of detection before and after correcting for the stellar variability using SCALPELS. Additionally, following Standing et al. (2022, 2023) we evaluate the effectiveness of our detection protocol by injecting low-mass planets with RV semi-amplitudes in the m s^{-1} level into our HARPS-N data and retrieving them with KIMA and SCALPELS.

3.1.1 Prior distribution

The models we fit for the two stars are defined by the priors given in Table 3. These priors are comparable to those described in Faria et al. (2020), but tailored for HD 166620 and HD 144579 independently.

We set the number of Keplerians N_p to be a free parameter with a uniform prior between 0 and 5. i.e. KIMA will try to fit a maximum of 5 Keplerians to the data simultaneously and independently of each other. For this search, we use a uniform distribution for ϕ , γ and ω , as there is no justification to favour any specific value within these parameter regimes. We use a Kumaraswamy distribution ($\alpha=0.867$ and $\beta=3.03$) (Kumaraswamy 1980) for planetary eccentricities, which favours lower values but still permits the exploration of higher eccentricities when the data requires it (Standing et al. 2022).

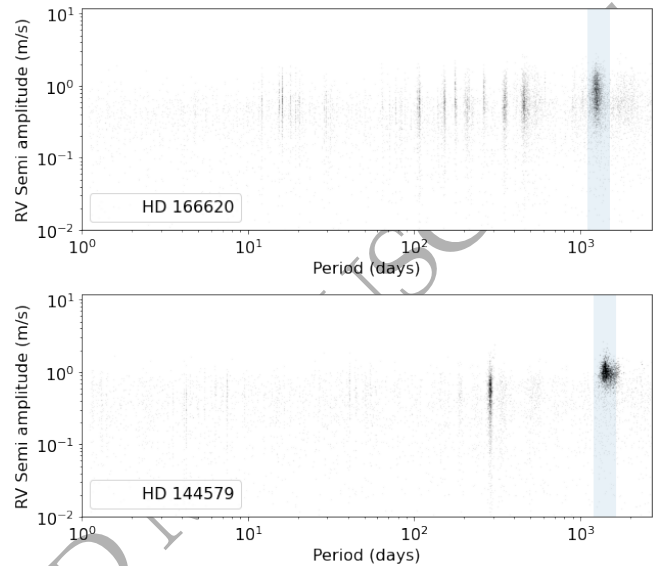


Figure 5. The KIMA joint posterior distribution of RV semi-amplitudes (in log scale) for HD 166620 and HD 144579 in log orbital period space are shown. Both stars exhibit an anomalous signal with a similar orbital period of 1400 days. The posterior orbital period space is populated with more over-density patches, which represent signs of additional signals, possibly planets, aliases and/or harmonics.

The values for shape parameters α and β are justified by the Beta distribution in Kipping (2013).

The stellar reflex orbital semi-amplitude (K) is the one parameter to sample with the highest sensitivity when establishing robust detection limits (Standing et al. 2022). We will describe this analysis in Section 4.1. In this regard, we use a modified Log-Uniform (Jeffreys) prior. The prior values given for K in Table 3 are obtained by carrying out a number of initial runs with uninformative priors, to establish a plausible upper limit. The model also accounted for the systemic velocity of the centre of mass of the system (γ : which corresponds to an RV offset measured by HARPS-N) and a jitter term (σ_{jit}) added in quadrature, to effectively represent the uncorrelated noise sources (uncorrelated at the timescale of the planet signals we are looking for) such as granulation and small night-to-night wavelength calibration errors. For σ_{jit} , we again use a Jeffreys prior. In order to account for the stellar activity, the SCALPELS **U**-vectors, representing the projection of line-shape changes onto the RV are included as decorrelation vectors (as in Anna John, Collier Cameron, & Wilson (2022)), using a Gaussian prior for their amplitude coefficients.

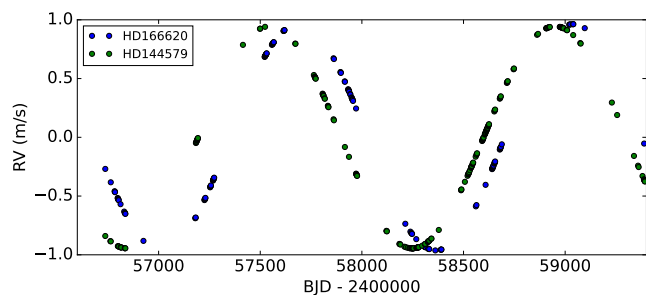


Figure 6. The orbits have been simulated using the best-fit parameters for the ~ 1400 d signal, observed in HD 166620 and HD 144579. The signals are nearly in phase (there is a small offset) and exhibit comparable RV variations of 95 cm s^{-1} . The phase offset possibly arises from the slightly different periods (1390 and 1420 days) of these signals.

3.2 1400 day periodicity

Figure 5 shows the resultant posterior distribution for the RV semi-amplitudes for HD 166620 and HD 144759, plotted against the posterior orbital periods. The analyses initially revealed a long-term periodicity of approximately 1400 days in both stars. We obtained meaningful fits for 1400-day sinusoids in both stars while decorrelating against the SCALPELS shape-driven variations. We found that these signals are also more or less in phase and share very similar sub- m s^{-1} (95 cm s^{-1}) semi-amplitudes (Figure 6). This suggested a systematic, rather than a dynamical origin. Therefore, we investigated further the nature and possible origin of these common periodic signals.

3.2.1 Instrumental zero-point subtraction

We searched for a common signal with a similar orbital period ($\sim 1400 \text{ d}$), phase, and semi-amplitude in 12 of the most intensively-observed RPS targets to gain clarity on whether or not this signal is common to all the RPS targets. From parallel analyses using YARARA (Cretignier et al. 2021) and SCALPELS (Collier Cameron et al. 2021), we discovered sporadic sub- m s^{-1} jumps in the observed RV data of most of the RPS targets, in a pattern that mimics a long-term periodicity. Further investigation revealed that these discontinuities in RV occurred at the times of maintenance interventions performed on the HARPS-N instrument. These were the periodic warm-ups performed on the detector as a temporary solution for the small leak in the detector cryostat (Dumusque et al. 2021). The detector cryostat was changed in October 2021. This solved the leak problem, but the jumps remain in observations made prior to this intervention.

We experimented with establishing a zero-point offset between the interventions and subtracting it from the target star RVs in order to account for these substantial instrumental interventions (cryostat warmups). A sample set of stars with long observation baselines, good seasonal sampling, and photon uncertainties of less than 1 m s^{-1} was selected for this purpose. Table A1 displays the list of stars (including the Sun) chosen using these criteria, the majority of which shared a comparable long-term periodicity. This enabled better quantification of the RV offsets between major interventions performed on the instrument. To prevent removing genuine signals from the data of HD 166620 and HD 144579 analysed in this paper, we removed those stars from the sample used to calculate the zero point offsets induced by detector warmups. i.e., we omitted HD 144579 from the sample set to calculate the zero points for HD 144579 and did the same for HD 166620.

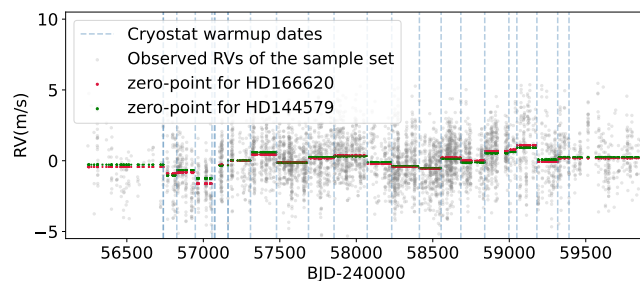


Figure 7. The estimated zero points in RVs between the epochs of cryostat warmups for HD 166620 and HD 144579 are shown by red and green points respectively. It is evident that when we analyse HD 166620 and HD 144579 individually, we find very similar values for the offsets. The grey points in the background represent the median-subtracted shift RVs of 11 other HARPS-N targets, corrected for activity variations using SCALPELS. The dashed grey vertical lines indicate the dates of instrumental interventions considered for the zero point estimations.

The zero-point estimation was performed as follows. The observed RVs of each star were first corrected for activity cycles by removing the SCALPELS shape component. We then removed the extreme outliers after subtracting the median RV from the activity-corrected shift RVs. We divided the data set into segments between the known dates of interventions and estimated the zero point as the median of the combined corrected RV data set in each segment. The floating chunk offset (FCO) technique, used to correct for night-to-night zero-point velocities by Hatzes (2014); Hatzes et al. (2011, 2010), served as the basis for chunk-by-chunk offset subtraction used in this study. The dates of the cryostat warmups that define the segment boundaries while estimating the zero points are represented by the vertical lines in Figure 7. Then, we simply subtract the estimated chunk-by-chunk zero-point (illustrated in red and green in Figure 7) from the observed RV data of the target stars for further investigation. The hence obtained BJD boundaries and zero-point offsets are given in Table A2. This finding serves as a powerful demonstration of the ability of HARPS-N to monitor such a sub- m s^{-1} level signal over a ten-year period.

3.3 Posterior analysis

3.3.1 HD 166620: The tale of a ‘quiet’ star

Figure 8 displays the RV semi-amplitude posteriors from KIMA, after they have been adjusted for the long-term instrumental systematics (using the zero-point subtraction method outlined in 3.2.1). The prevalent 1400 d signal that was caused by occasional cryostat warmups is no longer dominant (compare Figures 5 and 8).

We then directed our search for planet candidates in the offset-subtracted RVs using TWEAKS (described in section 3.1) that takes into consideration the instrumental shifts and stellar activity cycles in both the wavelength and time domains. Using the KIMA nested-sampling package (Faria et al. 2018) and a model with up to five unknown Keplerian signals, we performed a blind search of the radial velocities. With a sum of Keplerian functions from N_p orbiting planets, the algorithm models the RV time series while estimating the posterior distributions for all of the orbital parameters. With SCALPELS (Collier Cameron et al. 2021), which uses principal-component analysis of the autocorrelation function of the CCF, the time-domain activity-decorrelation vectors were computed. These

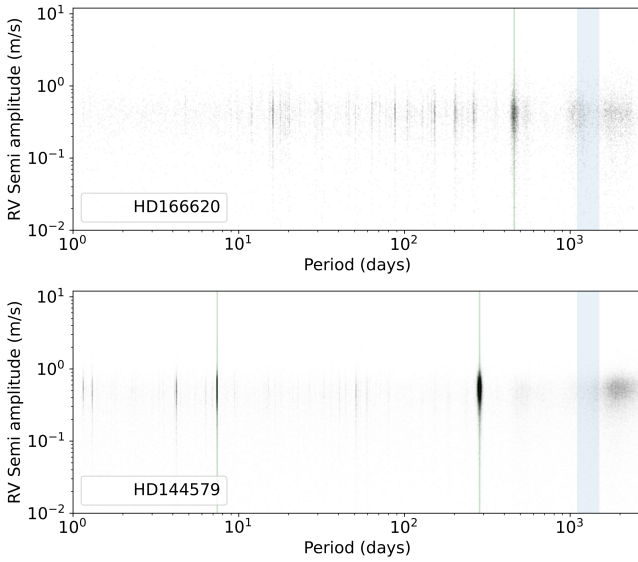


Figure 8. The joint posterior distributions from KIMA after subtracting the zero points estimated for the chunks of observations between major cryostat interventions, shown in Figure 7. The common long-term instrumental signal shown in Figure 5 with orbital period ~ 1400 d is hence modelled out. The displayed posteriors are obtained from more than 100000 independent trial models for HD 166620 (Top) and HD 144579 (Bottom).

vectors are then used as independent activity indicators for linear decorrelation in KIMA.

The top panel of Figure 8 displays the resultant joint posteriors for HD 166620 after the removal of the floating-chunk zero-point estimates between cryostat warm-ups and decorrelation against the SCALPELS decorrelation vectors. The posterior samples below the detection threshold appear as a background ‘fog’ following the priors for K and $\log P$.

A relatively well-defined signal appears at an orbital period of 460.59 d. Using a Gaussian mixture model based on Foreman-Mackey (2014), we computed the probabilities of posterior samples belonging to Gaussian foreground and background populations in $\log K$ and $\log P$, to exclude the posterior points coming from the uniform prior ‘fog’. This way, we isolated the foreground probability of the cluster and obtained a reliable estimation for the RV semi-amplitude for this signal as $0.41 \pm 0.10 \text{ m s}^{-1}$. If this signal is planetary, this would correspond to a planet with an upper $M \sin i$ mass limit of $3.34 \pm 0.85 M_{\oplus}$, when taking into account a stellar mass of $0.76 M_{\odot}$. This method increases the precision and accuracy of K compared to the standard measurement of the posterior in KIMA.

We conducted a False Inclusion Probability (FIP) analysis (Hara et al. 2022) in frequency space, with the bin size set to the Nyquist frequency resolution over the entire data duration. The Gaussian mixture model allowed us to calculate the probability of posteriors in the foreground, independently of the choice of frequency bin width. We first computed the True Inclusion Probability (TIP), as the fraction of all KIMA trial models (N_s) that contain planets whose periods fall in the frequency interval of interest, belonging to the foreground. In other words, TIP is the number of models for which the orbital period (P) and RV semi-amplitude (K) of a sampled planet fall within their respective foreground populations.

$$\text{TIP} = \frac{\sum_{i=1}^{N_s} \text{Pr}_i(\text{foreground})}{N_s} \quad (1)$$

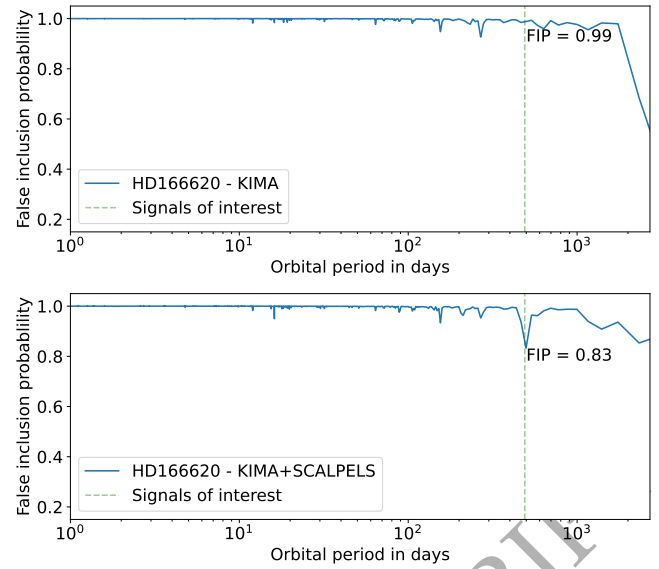


Figure 9. False Inclusion Probability (FIP) periodogram of HD 166620 showing the minimal values at orbital periods of potential detections. Top: FIP when no SCALPELS U-vectors are used for activity decorrelation in KIMA. Bottom: A more significant FIP value is observed at the same orbital periods when SCALPELS is included.

The FIP was then obtained as:

$$\text{FIP} = 1 - \text{TIP} \quad (2)$$

Hence, we simultaneously searched for multiple planets using a frequency window sliding across the full frequency range covered by the posterior distribution. A FIP value of 0.83 (Figure 9) was found at an orbital period of 460.59 d. This is not what we anticipate from a strong planetary signal detection, particularly for a star that is magnetically quiet. A FIP value of 0.83 indicates that only 17% of the models tested favoured a detection at this orbital period interval. To validate this, we performed several analyses by shuffling the observing seasons between years and found that the long-term signal at 460.59 d is comparable in its amplitude and its FIP to window-function artefacts appearing at other periods in the permuted datasets. We conclude that it too is most likely to be a window-function artefact.

3.3.2 HD 144579: A moderately active ‘lone’ star

Unlike in HD 166620, the joint posteriors from the HD 144579 RVs adjusted for long-term systematics show strong signals at 7.39 and 284.13 d (bottom panel of Figure 8) with RV semi-amplitudes of 0.50 ± 0.11 and $0.60 \pm 0.07 \text{ m s}^{-1}$. If these signals are planetary, they would correspond to planets with $M \sin i$ of $1.28 \pm 0.29 M_{\oplus}$ and $4.14 \pm 0.52 M_{\oplus}$, respectively, from the foreground posterior probability contained using a Gaussian mixture model. The obtained results favoured $\geq 4.5\sigma$ detections for both signals. However, based on the ratio of probabilities of consecutive values of the number of planets iterated in the KIMA model comparison, the sampler favoured zero planets (based on the 150 Bayes Factor threshold), as shown in Figure 12.

The False Inclusion Probabilities from the posteriors of HD 144579 are shown in Figure 10. In contrast to HD 166620, substantially lower FIPs are observed in HD 144579 at orbital periods of 7.39 and 284.13 d. Although there is a notable decrease in the FIP

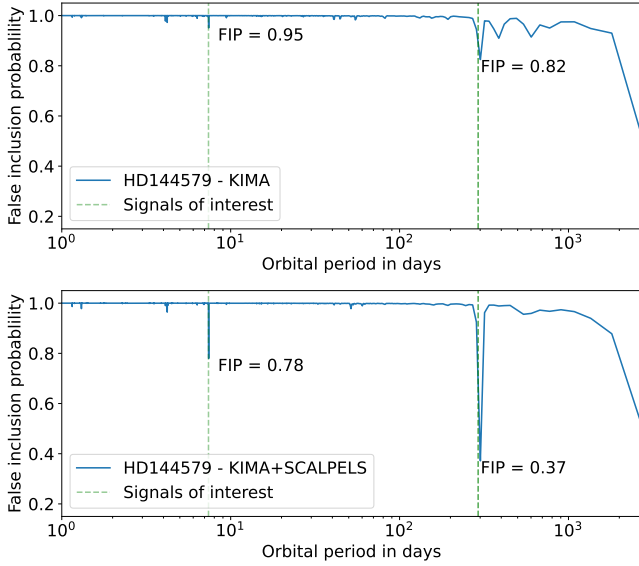


Figure 10. False Inclusion Probability (FIP) for HD 144579 before and after stellar activity decorrelation. The top and bottom panels are as in Figure 9. Here, the FIP periodogram of HD 144579 shows clearer detection of the signals of interest, when decorrelated against the SCALPELS U-vectors.

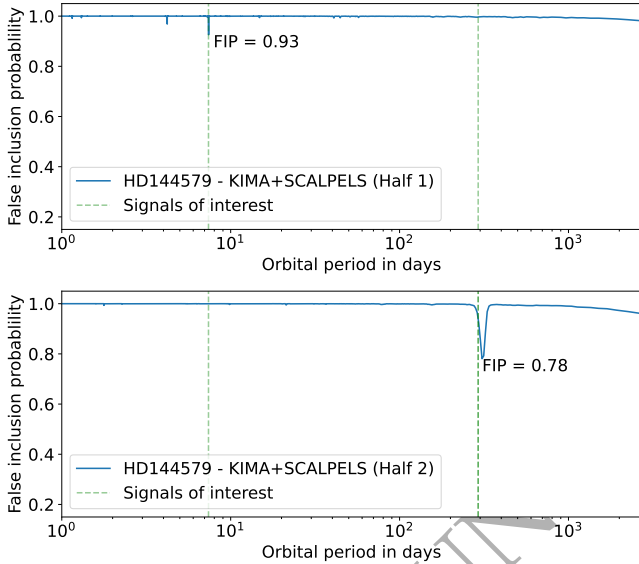


Figure 11. False Inclusion Probability (FIP) for HD 144579 before and after stellar activity decorrelation. The observations are now divided into two subsets to investigate the coherency of the signals of interest that appeared in the bottom panel of Figure 10. The FIP diagrams obtained for individual half subsets of the HD 144579 data are shown in the two panels. Each data subset covered an observation baseline of > 1500 days. Unlike any coherent signal, none of the signals of our interest was commonly detected in both data subsets. Hence, each of these signals could possibly be an artefact of the sampling pattern arising from individual seasons.

when SCALPELS decorrelation vectors are included in the model, the values are still not conclusive. Using the 284.13 d signal as an example, reporting a detection based on a FIP value of 0.37 (Figure 10) is analogous to boarding an aeroplane with only a 63% chance of making it to the destination.

Moreover, a number of ambiguous detections can be triggered at

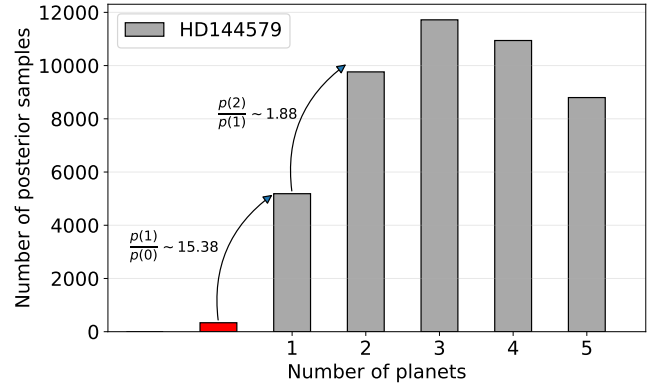


Figure 12. The posterior distribution for the number of planets N_p . The counts are the number of posterior samples in trial models with a specific number of planets. The probability ratios between models with 0, 1 and 2 planets are shown. The posterior distribution suggests (Figure 8) that there are likely to be planets present. The ‘confident detection’ criterion adapted by KIMA, however, takes into account the ratio of probabilities for successive values of N_p , which favours $N_p=0$, rather than depending solely on the probability values. The preferred model ($N_p=0$) appears underrepresented, because of the significantly larger parameter space for models with successively greater N_p . On the other hand, when a Keplerian was injected, KIMA favoured $N_p=1$ (Figure B2). A similar representation for HD 166620 is shown in Fig.B1.

Table 4. Derived posterior values for Signals of Interest (SOIs) from the KIMA+SCALPELS RV analysis detailed in Section 3.3. The FIP values given are obtained from the entire data. $M_p \sin i$ values are estimated considering if the SOIs are planetary.

Parameter (unit)	HD 166620 SOI.1	HD 144579 SOI.1	HD 144579 SOI.2
<i>Derived parameters from KIMA posteriors</i>			
P (d)	459.88 ± 11.05	7.39 ± 0.002	284.13 ± 4.16
K (m s^{-1})	0.41 ± 0.10	0.50 ± 0.11	0.60 ± 0.07
$M_p \sin i$ (M_\oplus)	3.34 ± 0.85	1.28 ± 0.29	4.14 ± 0.52
e	0.22	0.14	0.18
ω (deg)	2.97	3.48	3.01
T_0 (BJD-2400000)	57491.32641	58514.54291	58243.86417
FIP	0.83	0.78	0.37

spurious periods because of the sampling pattern and cross-talk between various aliases (e.g., Anna John, Collier Cameron, & Wilson 2022). To guard against this possibility, we analysed the system further by splitting the data train. The primary goal was to examine the individual data halves for any candidate signals. If the baseline covered by any individual subset is appropriate for sampling the signals of interest, then any coherent signal should be detectable independently in n number of data subsets, while the significance of detection depends on the size of the subset. There is a caveat, though. It is probable that the stellar activity can manifest differently in the subsets and the ability of TWEAKS to completely clean the time series could vary, leading to stronger signals in some cases, and weaker in others.

Figure 11 shows the FIPs obtained from the posteriors of individual half subsets. Contrary to what would be expected from any coherent planetary signal, neither of the signals of interest were consistently found in both of the two data subsets. Each of these signals could

Table 5. Bayesian evidence for models with and without SCALPELS vectors used for stellar activity decorrelation.

Star ID	KIMA ($\log \mathcal{Z}_1$)	KIMA + SCALPELS ($\log \mathcal{Z}_2$)	Bayes Factor $\mathcal{Z}_2/\mathcal{Z}_1$
HD 166620	-457.29	-453.65	38.09
HD 144579	-362.94	-357.56	217.02

therefore be an artefact of the sampling pattern resulting from a particular localised subset of the observations or the existing data may not be adequate to confirm these detections.

To validate this finding, we also performed some injection and recovery tests by injecting a Keplerian with an orbital period of 210.28 d and an RV semi-amplitude of 60 cm s^{-1} (comparable to the semi-amplitude of the 284 d signal). We then repeated the analysis and successfully recovered the injected Keplerian as independent marginal detections in individual half-data subsets (Figure B5). An injected signal with a slightly larger semi-amplitude (80 cm s^{-1}) was consistently better detected in the individual subsets as shown in Figure B6. We also experimented with injecting a short-period signal with $P = 5.12 \text{ d}$ which is closer to the 7.39 d signal and with an RV semi-amplitude $K = 60 \text{ cm s}^{-1}$. Unlike the long-period injected signal, this signal was only recovered in one of the half sets (Figure B3). On the other hand an 80 cm s^{-1} injected Keplerian was successfully recovered in individual subsets (Figure B4). This suggests that $K = 60 \text{ cm s}^{-1}$ is below the detection threshold for the data partitioning test at $P = 5.12 \text{ d}$. By the same token, it also suggests that the 7.39 d signal might not be entirely spurious. However, without confirmation in independent subsets of the data, we cannot be certain.

This provides a pragmatic estimate of the detection limit in the HARPS-N data to be somewhere between 60 and 80 cm s^{-1} , which is consistent with the formal estimation of detection threshold presented in Section 4.1. In addition, these simulations show that the data in hand are not sufficient enough to conclude the origin of the 7.39-day and 280-day signals.

3.4 Impact of stellar activity mitigation using SCALPELS, in the model selection

The difference in evidence (a.k.a Bayes factor) between a model with decorrelation performed against stellar activity using SCALPELS U-vectors and a model uncorrected for stellar activity is given in Table 5. The shape component comprising the corresponding U-vectors for individual stars are shown in orange colour in both panels of Figure 4. As anticipated for a star in a Maunder-minimum state with no identifiable magnetic activity cycles, the shape component obtained for HD 166620 looks essentially flat. However, SCALPELS effectively deals with the focus intervention that occurred in the HARPS-N instrument, which is seen as an evident jump in the shape time series, at around $\text{BJD} = 2456737$. This intervention was planned as the FWHM in all stars observed by HARPS-N was showing a significant drift, likely due to a change of instrument focus with time. A component was removed that was supposed to stabilize the focus over time but was clearly not performing as expected. This fixed the FWHM drift issue, however, a significant jump in FWHM, likely due to a PSF change, was observed after this intervention. This jump

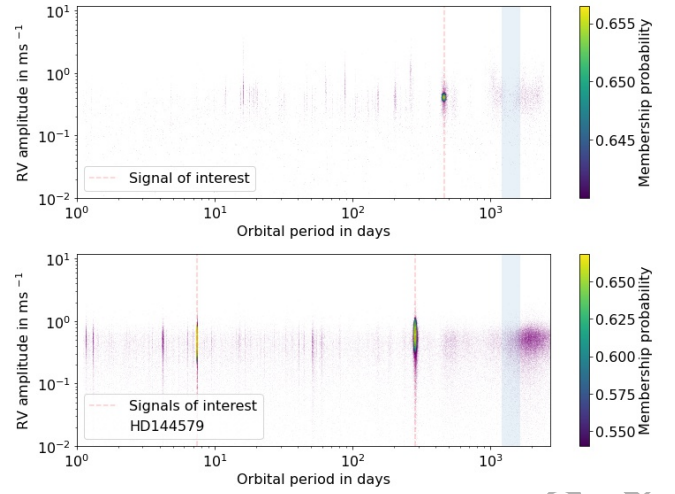


Figure 13. The purple points represent the entire KIMA joint posterior semi-amplitudes plotted against the orbital periods. The green to yellow points represent the posterior samples that correspond to the foreground probability of the candidate signals as determined by a Gaussian mixture model. A point is more likely to be in the foreground if it is more yellow. Top: HD 166620 Bottom: HD 144579.

has been efficiently tracked by some of the SCALPELS basis vectors (Figure D2 & D1).

As is evident from the shape component in the lower panel of Figure 4, SCALPELS identified a greater amplitude of shape-driven RV variation in HD 144579 than in HD 166620, owing to its greater intrinsic activity level. If we now take a look at Table 5, we can see how the Bayes Factor reflects the impact of shape corrections on each star. The Bayes factor provides a measure of how strongly one model is supported over the other by the data. We use the same Jeffreys' scale as a benchmark for our model evaluation. In accordance with Trotta (2008), we defined the threshold for "moderate" evidence at Bayes Factor = 12 and for "strong" evidence at Bayes Factor = 150. While SCALPELS decorrelation improves the model by a factor of 38 for HD 166620, HD 144579 shows a significant improvement of 217 times.

4 DISCUSSION

4.1 Detection limits

A distinct characteristic of a trans-dimensional nested sampler, such as KIMA, is that it will create a map of all signals that are consistent with the data when forced to explore higher N_p than is explicitly detected. Since those suggested signals are still technically undetected, this posterior can be used to generate a detection threshold (Standing et al. 2022). To establish the detection limits in HD 166620 and HD 144579, we typically adhere to the process described in Standing et al. (2022), where they use the Bayesian approach with KIMA to produce detection limits to answer the question "what is compatible with the data?". A well-sampled posterior is necessary for generating a reliable detection limit. The number of saves in KIMA was set to 200000 to obtain a minimum of 30000 effective posterior models. A robust detection limit is usually obtained after the removal of any candidate Keplerians present in the data. Here, we can directly compute the detection thresholds from the posteriors, since no valid detection can be established using the data presented.

The posterior is divided into log-spaced bins in the orbital period

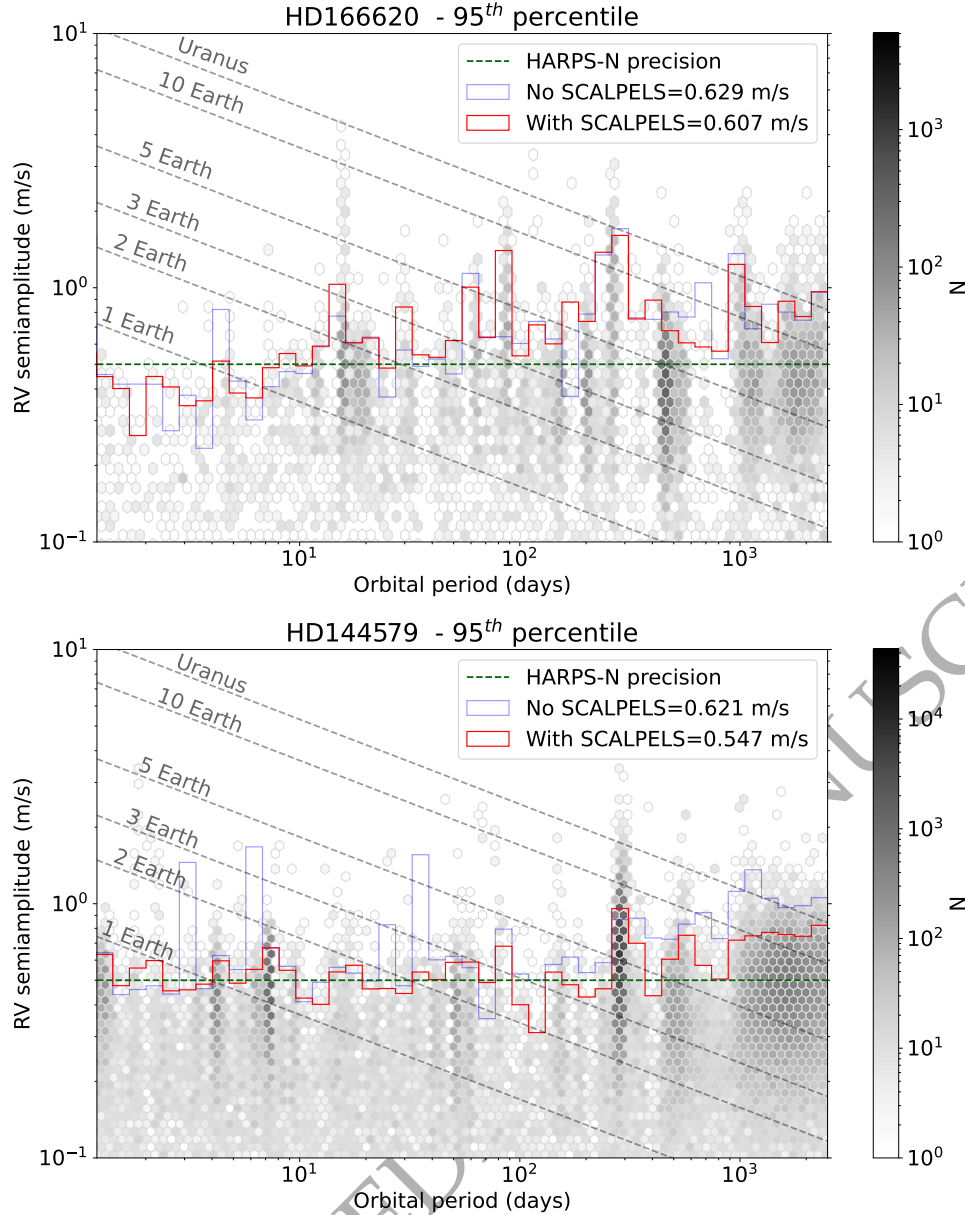


Figure 14. A greyscale hexbin plot representing the density of posterior samples derived from the KIMA+SCALPELS runs. Faded blue lines represent detection limits derived from a KIMA run without any decorrelation against the scalpels basis vectors. The solid red line shows the detection limit computed from a KIMA run corrected for stellar activity with SCALPELS. The grey dashed lines are contours of constant planet mass ($M \sin i$) for comparison.

(P) to calculate the detection limit. Within each bin, the 95th percentile of the semi-amplitude (K) distribution is then evaluated. The findings of our analysis of the detection limits for the HD 166620 and HD 144579 are shown in the top and bottom panels of Figure 14. The grey scale hexbins display the posterior sample density from the KIMA + SCALPELS trials on each target. The red lines enveloping the entire posterior space represent the calculated detection limit from a single run with decorrelation performed against the SCALPELS U-vectors. The no-SCALPELS version is used to compute the faded blue detection limit line.

Figures 9 and 10 depict the effect of SCALPELS correction in HD 166620 and HD 144579. While the window function artefact at $P \sim 460$ d in HD 166620 appears well below the detection threshold, the corresponding signals in HD 144579 at 7.39 and 284 d obstruct

the improvement in the detection limit despite being decorrelated against SCALPELS U-vectors.

4.2 Probing RV signals in the sub-m/s regime

Doing the stellar activity decorrelation in the wavelength domain using SCALPELS also has an evident impact in bringing down the detection limit more towards the sub m s^{-1} regime. In Figure 14, the detection limits calculated with and without SCALPELS decorrelation are shown. Although the impact varies slightly across various period regimes, SCALPELS decorrelation allows for better probing in the low-mass regime in both stars. This is in line with a recent study by Sairam & Triaud (2022), in which they show how the activity cycle affects the detection efficiency of planets in the solar case.

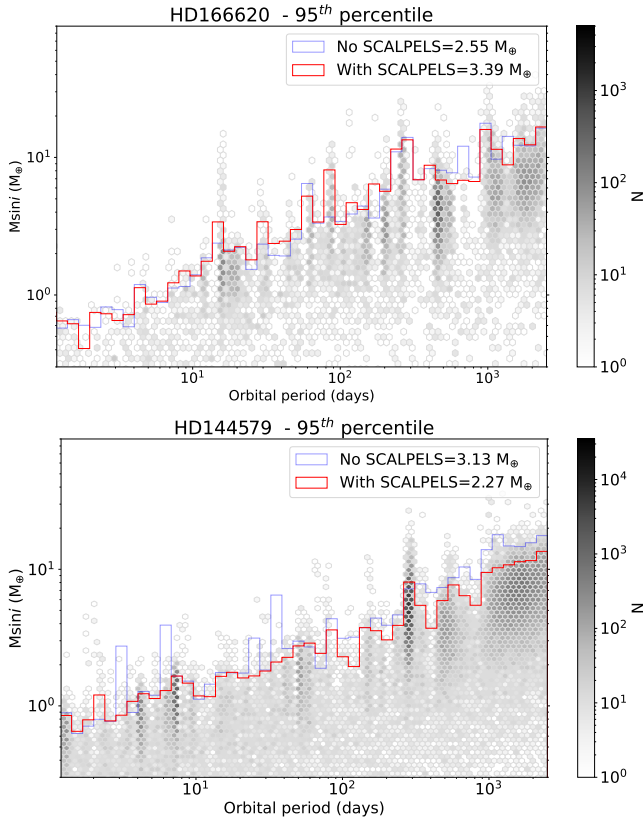


Figure 15. Upper detection limits in the $M \sin i$ versus orbital period space. The blue and red stepped histograms respectively show the 95th percentile detection limit before and after decorrelating against SCALPELS U-vectors.

They found that the likelihood of finding a low-mass planet falls by an order of magnitude during the solar activity maximum. Therefore, they suggest a more efficient observational strategy based on a forecast of stellar activity that would enhance the identification of Earth counterparts and their analysis. Our study shows that we could detect smaller planets, when we correct for the stellar activity and instrumental shifts using SCALPELS, despite the activity state of the target star. However, the level of magnetic activity in the individual stars is reflected in the improvement in the ability to detect low-mass planets on the application of adequate stellar activity correction.

While the detection limit does not show significant enhancement in HD 166620, the mean detectability in HD 144579 improves from 0.62 to 0.54 m s^{-1} after correcting for the spectral-line shape changes. The most striking improvement is observed in the long-period regime, where the detection threshold consistently improves at periods longer than 300 days, reducing the detection threshold from $\sim 1.2 \text{ m s}^{-1}$ to $\sim 0.6 \text{ m s}^{-1}$ at periods around 1000 days. In the era of upcoming missions like PLATO and HARPS3, striving for Earth-like planets in an Earth-like orbit around a Sun-like star, these findings are of high importance. It is evident from Figure 14 that we are sensitive to a wide variety of Super-Earths ($5\text{--}10M_{\oplus}$) in the entire orbital period range ($P < 2800 \text{ d}$) spanned by the posterior space. Moreover, we can detect planets with masses as low as $1\text{--}3M_{\oplus}$ ($P < 300 \text{ d}$), and more massive planets with $5M_{\oplus}$ up to $P = 1000 \text{ d}$. The mean detection threshold computed after correcting for stellar activity falls around 60 cm s^{-1} in HD 166620 and 54 cm s^{-1} in HD 144579, approaching the HARPS-N precision limit imposed by wavelength solution (Dumusque et al. 2021), which,

without considering instrument intervention, should provide a similar long-term calibration precision for HARPS-N.

Figure 15 displays the same detection limit figure translated into a $M \sin i$ -orbital period space. It is now easily readable that under $P = 10 \text{ d}$, we are able to detect significantly lower-mass planets in HD 166620 and HD 144579 with masses smaller than Earth.

4.3 Areas for further development

Given that TWEAKS does not completely model the rotational modulation of the star, there is room for further advancement. Anna John, Collier Cameron, & Wilson (2022) found that some shift-like patterns evade SCALPELS analysis. They attribute the origin of these patterns to stellar rotation. We tried to use a GP to model any residual rotationally modulated signals. We were unable to proceed, as the typical Gaussian Process hyperparameters change considerably over the course of a cycle, making it more difficult to include in studies with long-term follow-ups like RPS. In addition, when we attempt to model the data using a single quasi-periodic GP, injected planetary signals with orbital periods longer than the active-region lifetime parameter are modelled as long-term trends, becoming attenuated. A multi-variate GP using the SCALPELS basis vectors as activity indicators could be used to optimally resolve this (e.g., Rajpaul et al. 2015; Barragán et al. 2022). However, this is beyond the scope of the current investigation.

5 CONCLUSIONS

In this study, we report a thorough investigation of the HARPS-N spectroscopic data of bright K and G dwarfs HD 166620 and HD 144579, neither of which is known to host a planetary companion. We examined these stellar systems using TWEAKS, that combines wavelength-domain and time-domain stellar activity mitigation using the KIMA and SCALPELS. The major objective was to search for planetary reflex-motion and investigate the impact of stellar activity mitigation on the capability of detecting low-mass planets, in stars with different levels of magnetic activity. We found no significant detections in either of the stars. We also ruled out the possibility of being misled by erroneous signals from sampling patterns by performing data splitting exercises and validating using injection recovery tests. The injection recovery tests also showed that the data in hand might be sufficient enough to conclude the origin of the 7.39-day and 280-day signals. The data partitioning test is something we would advise as a standard procedure to confirm the coherency of a signal before making any detections public. More rigorous coherency tests have also been proposed by Gregory (2016) and Hara et al. (2022).

Additionally, we provide an estimate for the detection limits on the HARPS-N radial velocity data using posteriors from KIMA diffusive nested sampler. We demonstrate the varied impact of stellar activity on the detection efficiency of planets in HD 166620 which is in the Maunder minimum state and HD 144579 which is moderately active. The likelihood of finding a low-mass planet increases noticeably across a wide period range when the inherent star variability is corrected for using SCALPELS U-vectors. The 54 cm s^{-1} detection threshold achieved based on the aforementioned decorrelations brings us closer to the known calibration precision value provided by the HARPS-N instrument (50 cm s^{-1}) (Dumusque et al. 2021).

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DATA AVAILABILITY

The HARPS-N data products used for the analyses are from the Data & Analysis Center for Exoplanets (DACE) database dedicated to the visualization, exchange and analysis of extrasolar planets' data. The PYTHON codes and notebooks used to generate the results and diagrams in this paper will be made available through the University of St Andrews Research Portal.

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APPENDIX A: INSTRUMENTAL ZERO-POINT CORRECTION

Table A1 shows the information about the 12 stars included in the

Table A1. RPS targets used in the sample dataset to monitor the 1400-day periodicity and calculate the zero points. Note: HD 166620 is taken out from the list when estimating zero points for itself. Similarly, for HD 144579 as well.

Star	No. of observations	After nightly binning	Time span (days)
HD 10476	1062	257	4108
HD 122064	709	280	4187
HD 127334	1432	388	4023
HD 128165	473	268	3674
HD 144579	888	257	3646
HD 158633	626	167	3301
HD 166620	947	318	4080
HD 32147	505	139	4091
HD 3651	720	154	3558
HD 4628	1540	379	4099
HD 62613	527	157	3976
Sun	857		

Table A2. BJD boundaries and RV zero-point offsets calculated from the full sample including both HD 166620 and HD 144579. The 'nan' values for offsets represent the intervals where no observations were obtained.

Date of intervention	RV zeropoints
2456738.50000	-0.2805
2456826.50000	-0.9587
2456947.50000	-0.7384
2457056.50000	-1.6168
2457072.50000	nan
2457076.50000	nan
2457161.50000	-0.2981
2457308.50000	nan
2457478.50000	-0.1927
2457687.50000	0.4363
2457854.50000	-0.0994
2458071.50000	0.1848
2458231.50000	0.4039
2458412.50000	-0.2475
2458554.50000	-0.4048
2458683.50000	-0.5337
2458839.50000	0.2337
2458997.50000	0.0001
2459049.50000	0.6544
2459180.50000	0.7807
2459316.50000	1.0554
2459390.50000	-0.0803

sample set used to estimate the RV zero-point between the cryostat warmups described in Section 3.2.1. The epochs of cryostat warmups and the consequently obtained RV zero points between the intervals of these instrumental interventions are given in Table A2.

APPENDIX B: INJECTION & RECOVERY TESTS

The FIP periodograms from the injection recovery tests, performed with injected Keplerians of varied orbital periods and RV semi-amplitudes are shown in Figures B3, B4, B5 and B6. As discussed in Section 3.3.2, we performed 4 injection and recovery tests.

- 1) Injecting a signal with $K=60$ cm/s and $P=5.12$ days (Figure B3), closer to the signal of interest at $P=7.34$ days.
- 2) Injecting a signal with $K=80$ cm/s and $P=5.12$ days (Figure B4)
- 3) Injecting a signal with $K=60$ cm/s and $P=210.28$ days (Figure B5), closer to the second signal of interest at $P=285$ days.

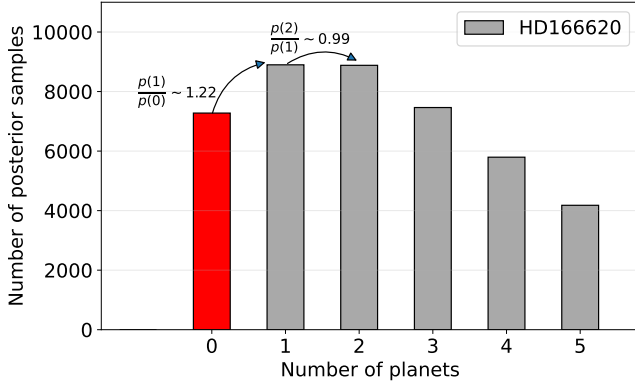


Figure B1. The posterior distribution for HD 166620, for the number of planets N_p . The counts are the number of posterior samples in trial models with a specific number of planets. The probability ratios between models with 0,1 and 2 planets are shown. The posterior distribution suggests that there are likely to be ‘zero’ planets present.

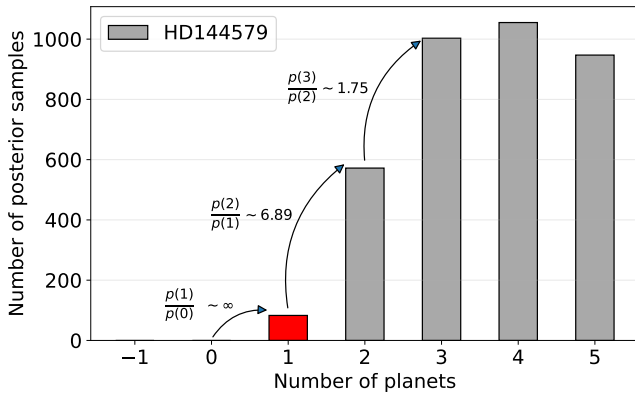


Figure B2. A Keplerian with $P=210.28$ d and $K=0.8$ m s $^{-1}$ is injected into the data. The ratio of probabilities for successive values of N_p now favours $N_p=1$.

4) Injecting a signal with $K=80$ cm/s and $P=210.28$ days (Figure B6). These injection & recovery tests helped us to validate the conclusions about the incoherency of the signals of interest, as the injected signals appeared independently in individual data halves (weak though in the case of 60 cm/s signals), unlike the SOIs. The strength of detection as shown by the FIPs (in B5 and B6) improved dramatically from 0.42 to 0.18 when the semi-amplitude of the injected signal was increased from 60 cm/s to 80 cm/s, which aided us in estimating the pragmatic detection limits to lie between 60 and 80 cm/s.

The FIP appears undoubtedly higher in the individual halves than the full data set (in both 60 cm/s and 80 cm/s cases). However, the idea here was to investigate if the injected signals show up independently in individual data halves with robustly defined orbital periods. The notable differences in the FIP in the full data also point towards the fact that doubling the number of observations improves the significance of detection drastically. This will also be important in planning observations in order to make good use of telescope time.

APPENDIX C: KIMA PLANET MODELS

The KIMA transdimensional nested sampling algorithm also provides a posterior distribution of the number of planets favoured by the data based on the ratios of Bayes Factor (or evidence) between suc-

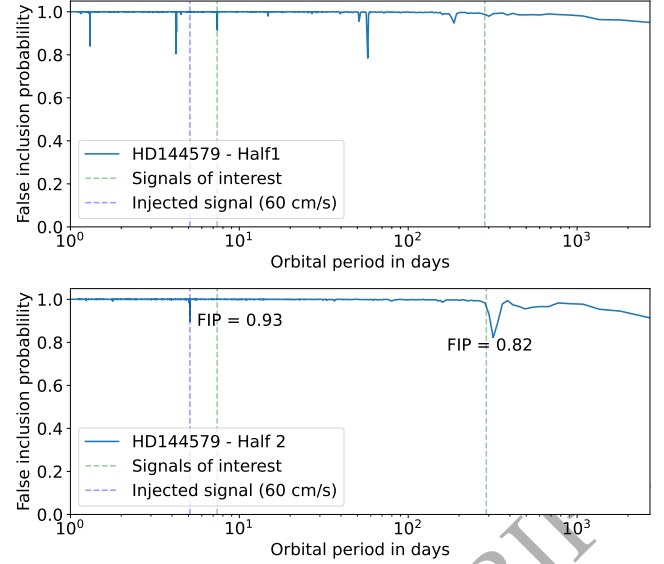


Figure B3. A Keplerian with $P=5.12$ d and $K=0.6$ m s $^{-1}$ is injected into the data. The FIP diagrams obtained for individual half subsets of the HD 144579 data are shown in the two panels. The injected Keplerian is only detected in one of the data sets.

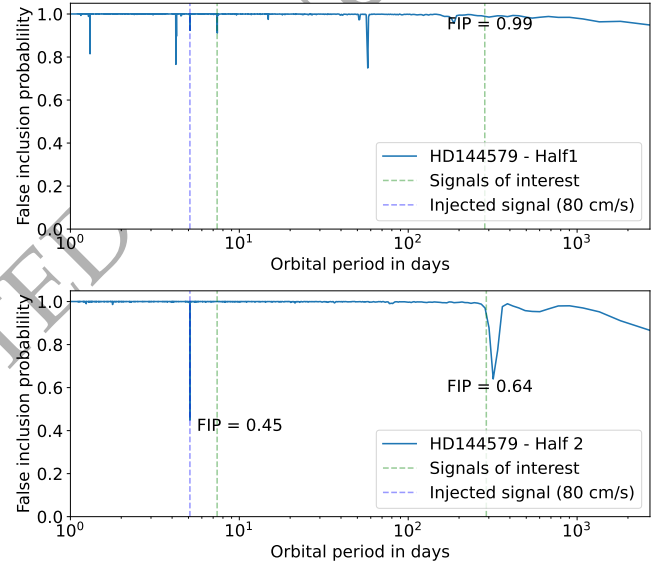


Figure B4. A Keplerian with $P=5.12$ d and $K=0.8$ m s $^{-1}$ is injected into the data. The FIP diagrams obtained for individual half subsets of the HD 144579 data are shown in the two panels. The injected Keplerian is now detected in both data sets independently, strongly in one half and marginally in the other half. This can be due to the cross talk of the injected signal between the nearby 7.34 day signal and its strong 1-day alias possibly arising from the sampling pattern of the first half data set.

cessive models with different numbers of keplerians. A model with N_p planets is considered to be the most favoured (strongly detected) when this ratio $Z_{N_p}/Z_{N_p-1} > 150$, and moderately detected when > 38 , based on the Jeffreys criteria. Figure B1 shows this representation for HD 166620, while the complementary Figure for HD 144579 is discussed in Figure 12, both of which show the supported zero-planet models. Figure B2 on the other hand shows the favoured ‘1-planet’

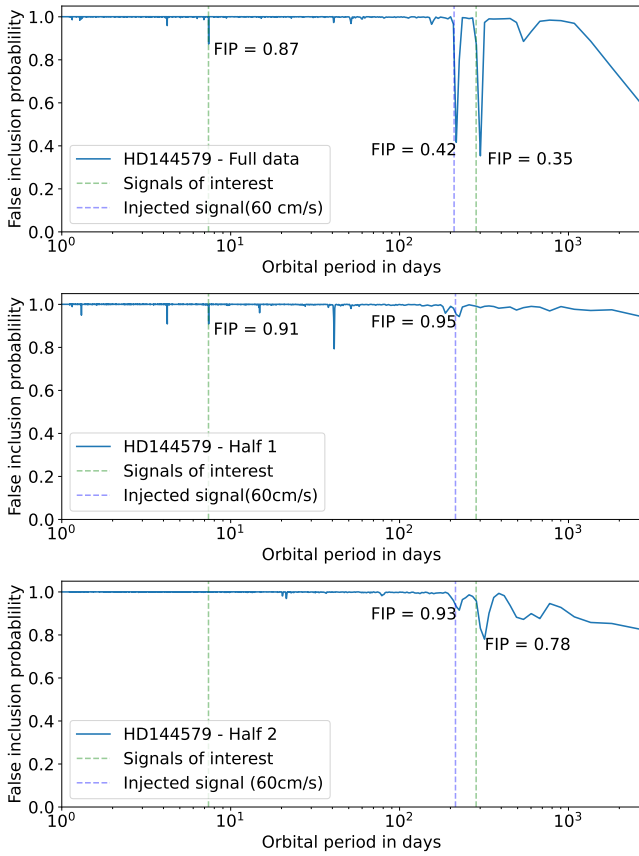


Figure B5. A Keplerian with $P=210.28$ d and $K=0.6$ m s⁻¹ is injected to the data. The observations are then divided into two subsets to investigate how a coherent signal appears in the individual halves. The FIP diagrams obtained for individual half subsets of the HD 144579 data are shown in the two panels. Unlike the signals of interest at $P=7.39$ and 284 d, the injected Keplerian is marginally detected in both data sets independently.

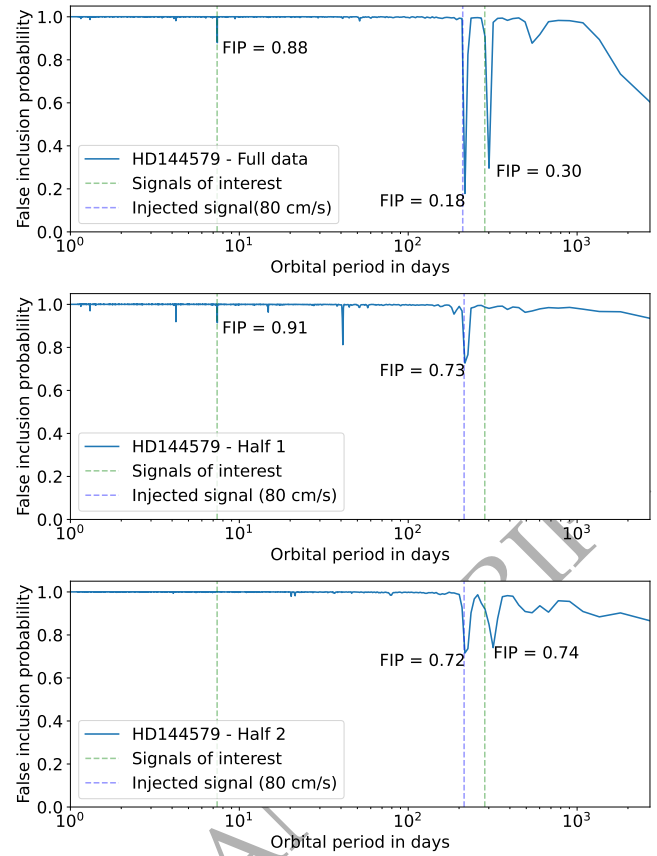


Figure B6. A Keplerian with $P=210.28$ d and $K=0.8$ m s⁻¹ is injected to the data. The FIP diagrams obtained for individual half subsets of the HD 144579 data are shown in the two panels. The injected Keplerian is now better detected in both data sets independently.

model on the injection of the 80 cm/s signal in HD 144579, pointing towards the detection threshold obtained in Section 4.1.

APPENDIX D: SCALPELS U-VECTORS & ACTIVITY INDICATORS

We also examined the first 5 leading principal components that contributed significantly in modelling the stellar activity (orange timeseries shown in Figure 4), to understand how efficiently the Principal Component Analysis of the ACF correlate with the known activity proxies such as FWHM, area and Bisector span. The corner plots showing the correlation between the SCALPELS U-vectors and traditional activity indicators are shown for HD 166620 and HD 144579 in Figure D2 and D1 respectively.

In HD 166620, the first and second principal components (U_1 & U_2) effectively trace down the FWHM and area of the CCF respectively. In HD 144579, the first principal component (U_1) exhibits a strong resemblance to the variability of CCF area. Another correlation is found between the second principal component (U_2) and the FWHM. This indicates that the changes in the width of the profile are also reliably considered while modelling the shape-driven component.

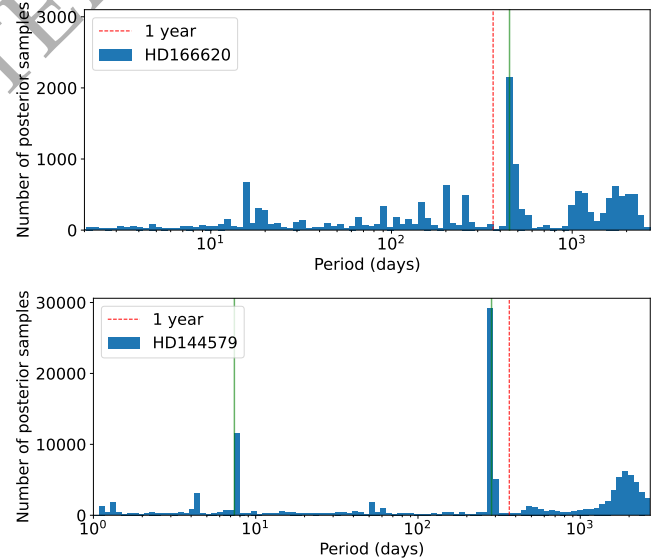


Figure B7. Histograms showing the number of independent posterior samples for each star. The signals of interest are marked with green vertical lines.

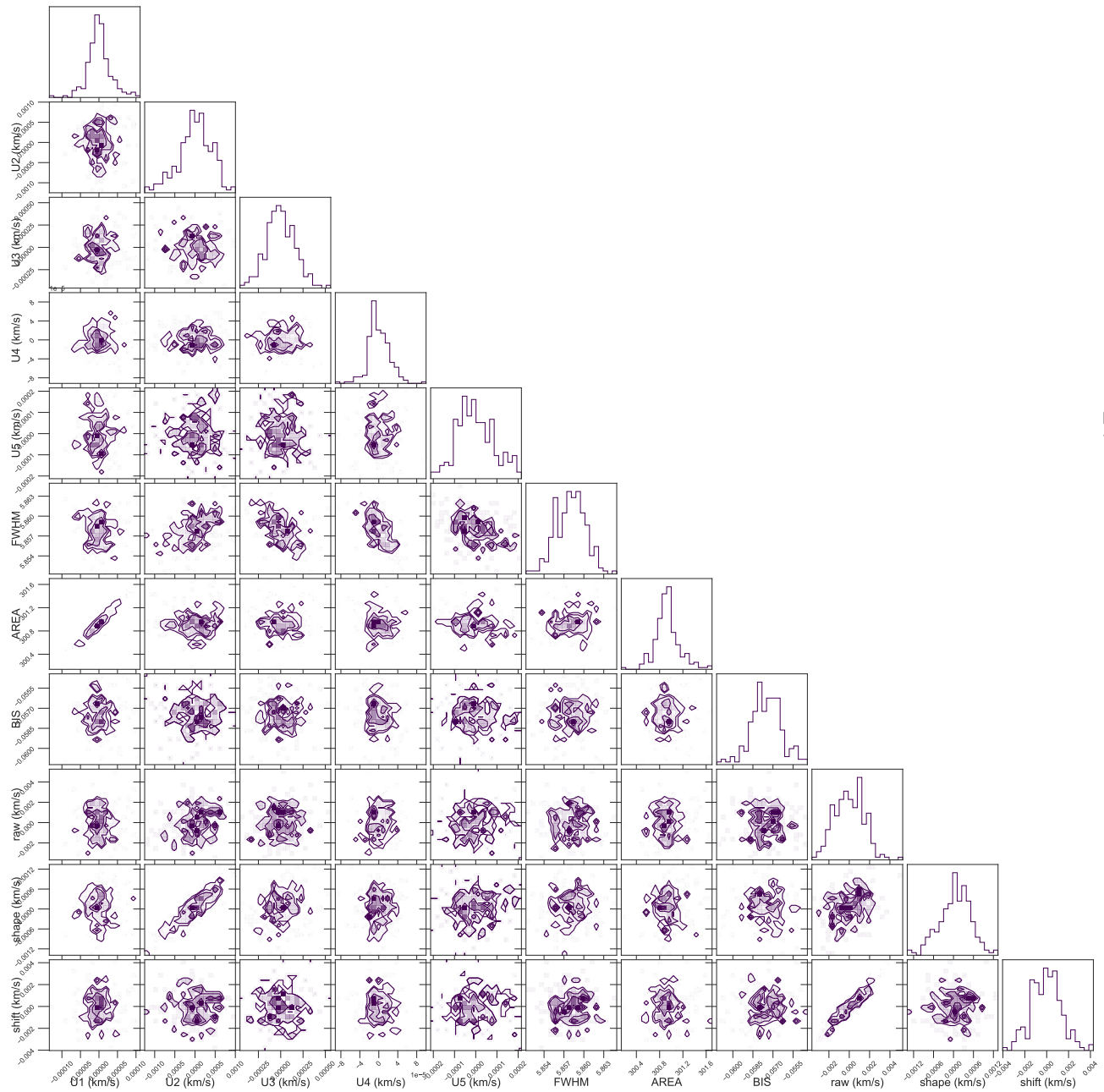


Figure D1. The correlations plots for HD144579 showing the leading U-vector components of the residual CCF of RV time series and the traditional activity indicators such as FWHM, area and Bisector span. U_2 is the prominent contributor to the entire shape component which shows a weak correlation with the FWHM, while the U_1 component is strongly correlated with the area. U_3 shows an anticorrelation with the FWHM.

ORIGINAL

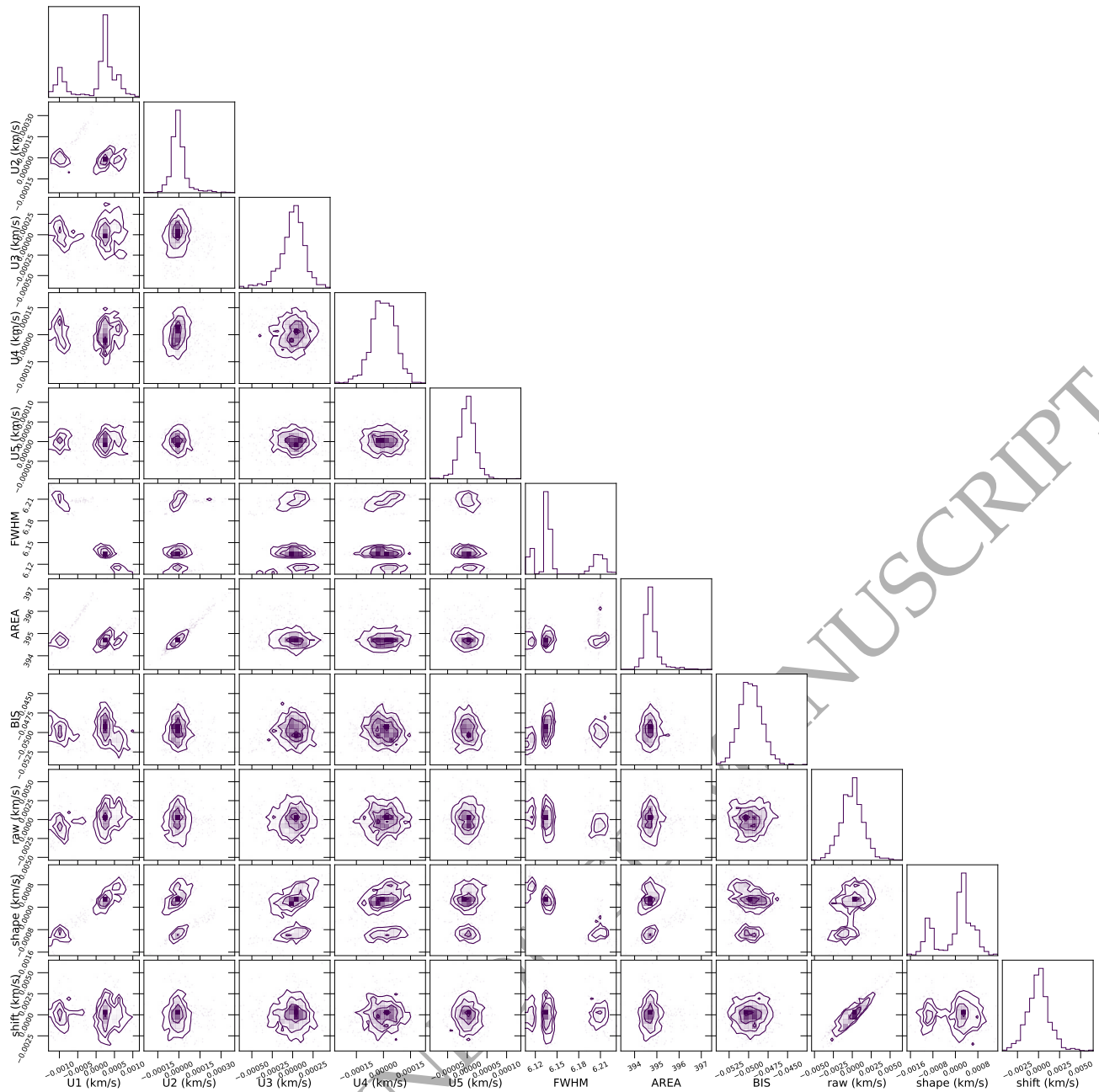


Figure D2. The correlations plots for HD166620 showing the leading U -vector components of the residual CCF of RV time series and the traditional activity indicators such as FWHM, area and Bisector span. U_1 is the prominent contributor to the entire shape component which shows a strong anti-correlation with the FWHM, while the U_2 component is strongly correlated with the FWHM.