A holistic and probabilistic approach to the ground-based and spaceborne data of HAT-P-19 system

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1 INTRODUCTION

Transiting exoplanets are of interest for several reasons. First of all, many physical properties of the planet and the host star can be directly measured or derived based on a minimum set of assumptions and the established theory of stellar evolution. Furthermore, variations in their transit timings can be indicative of an unstable orbit or additional bodies gravitationally bound to the system. In addition, planet formation and migration mechanisms of close-in giant planets can be tested based on a sample space containing well-characterized planets that have been discovered frequently in transit surveys due to the biases of the technique. As of now, even atmospheric constituents and properties of a handful of such planets have been constrained. There is currently a big effort to probe the atmospheric constituents and properties of a handful of such planets more accurately and precisely to carry out a population analysis of the planets within this mass regime to understand their formation, orbital, and physical properties of the host star.

In addition, planet formation and migration mechanisms of close-in giant planets can be tested based on a sample space containing well-characterized planets that have been discovered frequently in transit surveys due to the biases of the technique. As of now, even atmospheric constituents and properties of a handful of such planets have been constrained. There is currently a big effort to probe the atmospheres of more and more of these exoplanets, thanks to new highly performing instruments. This kind of investigations should become more accurate with the JWST and future extremely large class telescopes. Therefore, it is very worthwhile to monitor these systems frequently, analyse their archival data, and study them in more detail. From this perspective, the aim of this work is to refine the physical parameters of the planetary system HAT-P-19, mainly based on the analysis of (i) new and precise photometric follow-up observations and (ii) high resolution spectra from the Keck/HIRES archive. We also investigate possible transit timing variations (TTVs) of the hot-Saturn HAT-P-19 b by analysing our own observations, the light curves recorded by the Transiting Exoplanets Survey Satellite (TESS) (Ricker et al. 2015) and Spitzer telescopes, as well as other ground-based photometric observations of professional and amateur observers.

HAT-P-19 b is a low-density ($\rho_p = 0.293$ g cm$^{-3}$) Saturn-mass ($M_p = 0.95 M_{\text{Saturn}}$) exoplanet orbiting a metal-rich ([Fe/H] = +0.243 dex) star, on a $P_{\text{orb}} = 4.01$ d orbit with an eccentricity consistent with zero ($<0.024$; 1σ upper limit; Bonomo 2017). Its host star is a typical K-type main-sequence star with $T_{\text{eff}} \sim 4970$ K, and log $g = 4.567$. Most of the confirmed transiting planets with a mass roughly equal to that of Saturn have lower densities than that of the pure water, orbiting their host stars with diverse physical and orbital properties. Therefore, it is crucial to determine the parameters of such systems accurately and precisely to carry out a population analysis of the planets within this mass regime to understand their formation, orbital evolution, and the correlations between their parameters and that of their host stars.

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High iron abundance in the host star’s atmosphere, in contrast of the low density of its planet makes it a contradictory example for the correlation suggested to be existing between the core mass of the planet and the metallicity of the star in this mass regime (Hartman et al. 2011) which was inspired by a similar correlation between the radius of the planet and the metallicity of the host star for slightly more massive (0.4–0.7 $M_{\text{jup}}$) planets (Enoch et al. 2011). This sub-group of gas giants with intermediate masses ($0.1 M_{\text{jup}} < M_p < 0.4 M_{\text{jup}}$), low densities ($\rho_p < 2.00 \text{ g cm}^{-3}$), and relatively small orbits (a < 1.0 au) are called hot-Saturns. Together with their larger cousins, called hot-Jupiters, they pose a strong challenge for core-accretion scenario, requiring a migration mechanism to drag the planets from where they formed to where they are observed now. However, a relatively recent study has asserted in situ formation within core-accretion scenario as a potential mechanism for the formation of hot-Jupiters and Saturns (Bailey & Batygin 2018). The observed radius anomaly in these hot and bloated Saturns can be explained considering the incident flux coming from their host stars observed radius anomaly in these hot and bloated Saturns can be computed from theoretical models of Fortney et al. (2007) is still below the limits expected from the high metal content in such proximity of supersolar metallicity stars (Feltzing & Gonzalez 2001), where they might be claimed to have formed. In the case of in situ formation, kinetic heating (Guillot & Showman 2002) might be the mechanism that can unravel the story behind the inflation of such planets orbiting metal-rich stars. HAT-P-19 b is an important member of this class of planets, whose cores are expected to be enriched in metal content, therefore, should have been larger if they form where they are observed. If they had formed outside the snow-line and have migrated inwards, then the reason behind the inflation of their atmosphere, as well as the type of their migration history, are the questions to answer.

HAT-P-19 b’s extended radius ($1.064 R_{\text{p,jup}}$) for its mass ($0.284 M_{\text{jup}}$) also makes this hot and inflated Saturn-mass planet an outstanding target for transmission spectroscopy. So far, XO-2 b is the only planet orbiting a metal-rich star ([Fe/H] ~ 0.45, Teske et al. 2015) with the detection of both Na and K in its atmosphere (Sing et al. 2011; Pearson et al. 2019). Another hot-Saturn type planet WASP-49 b was also reported to have aerosol constraints (Cubillos et al. 2017) and neutral sodium at its high altitudes (Wytenbach et al. 2017) of its extended atmosphere with cloud decks (Lendl et al. 2016). Only Mallonn et al. (2015) attempted at obtaining a transmission spectrum of HAT-P-19 with the OSIRIS spectrograph at the Gran Telescopio Canarias. They did not find any trace of an additional absorption at any wavelength or any slope in their differential spectrophotometric search. With the same instrument, Sing et al. (2011) found a trace of potassium in the atmosphere of XO-2b. Nevertheless, HAT-P-19 b is still a promising target in terms of its potential of bearing heavy elements such as sodium and potassium, which may have escaped detection because of the pressure broadening in the planetary atmosphere (Mallonn et al. 2015). Whether such elements are found or not, this will provide another evidence for or against a suspected correlation between heavy element content in exoplanet atmospheres and planetary mass (Nikolov et al. 2018). These questions about the atmospheric content of this interesting planet, which is suggested as one of the prime targets to be observed with the James Webb Space Telescope (JWST) (Moliere et al. 2017), can be answered with the transmission spectroscopy observations from the space in the future. Therefore, determination of the planet and host star properties as a result of a detailed analysis of existing and new data with a holistic approach making use of recently developed analysis codes and global modelling techniques will help in characterization of its atmosphere too. Refining its ephemeris information is also crucial for planning future observations of the target with JWST and ground-based telescopes.

The system is also of particular interest due to the linear trend observed in its radial velocity residuals (Hartman et al. 2011), potentially indicating a gravitationally bound, yet unseen companion perturbing its orbit (Hartman et al. 2011; Seeliger et al. 2015) and/or causing the arrival times of the light from the system to the observer to change continuously (known as the Light Time Effect, LiTE). Nevertheless, studies of its TTVs (Seeliger et al. 2015; Maciejewski et al. 2018) have been inconclusive so far for the existence of such a potential third body. Only a few studies attempted to observe the target with high photometric precision to derive its parameters as well as its transit mid-times since its discovery back in 2011 (Hartman et al. 2011; Seeliger et al. 2015; Maciejewski et al. 2018). However, the system has been observed many times by amateur observers, as a result of which, at least a dozen moderate-quality light curves have been accumulated in the public archive of the Exoplanet Transit Database (ETD), having the potential to be used in a TTV analysis. Transits of HAT-P-19 b have also been observed with TESS in the cycle-2 during sector 17.

We observed the target several times with the 1 meter Turkish telescope T100, located in the Bakürittepe campus of the TÜBİTAK National Observatory of Turkey (TUG) at an altitude of 2500 m above sea-level, near the south coast of the country; and achieved very high photometric precision thanks to the well-established telescope defocussing technique (Southworth et al. 2009; Baştürk et al. 2015). We made use of the same technique during two very precise transit observations of HAT-P-19 b with the Zeiss 1.23 m telescope in Calar Alto Observatory (CAHA), which is located in Spain at an altitude of 2170 m. We analysed the most precise transit light curves of our own as well as that from the literature (Hartman et al. 2011) acquired by ground-based facilities, the transit observations of TESS, and two occultation observations by Spitzer with the state-of-the-art second version of the exofast software package (Eastman 2017, 2019), derived the global parameters of the system making use of the radial velocity observations by Hartman et al. (2011), and the atmospheric properties of the host star that we obtained from our own analysis of the Keck/HIRES archival spectra with primarily the ispec software package (Blanco-Cuaresma et al. 2014; Blanco-Cuaresma 2019). We made use of the brightness of the target in different broad-band filters and fit its spectral energy distribution based on its precise distance value thanks to Gaia mission to constrain the radius of its host star (Gaia Collaboration 2016, 2018). In addition, we collected all the light curves obtained so far by a number of observers around the world, all of which we corrected for the barycentre of our Solar system, measured mid-transit times in Dynamical Barycentric Julian Days (BJD-TDB), updated the ephemeris information for the system, established and analysed the TTV diagram. We present the photometric and spectroscopic data that we used and the details of the data reduction procedure in Section 2. We provide the information on data analysis and present the parameters of the system as a result of the global modelling of the data, and the TTVs in Section 3. Finally we discuss the importance of our findings in the context of the hot-Saturn type planets, their formation, orbital evolution, and inflation of their atmospheres, as well as the potential reasons behind the observed TTVs in Section 4.

1http://var2.astro.cz/ETD/
2 OBSERVATIONS AND DATA REDUCTION

2.1 Photometric data

2.1.1 TUG T100 light curves

We observed eight transits of HAT-P-19 b between July 2014 and August 2019 with the 1 m Turkish Telescope TUG T100 and the high quality, cryo-cooler SI 1100 CCD with 4096 × 4096 pixels, which gives a field of view of 21.5 arcmin × 21.5 arcmin at a plate scale of 0.32 arcsec per pixel. We observed through Cousins R, as this filter gave the highest count rate for photons from the target star, and without binning or windowing. The telescope was defocused in order to increase the photometric precision of the observations (Southworth et al. 2009) and using our standard approach for this and other telescopes (Mancini et al. 2014). Although we used autoguiding, we found that the centroid of the target star drifted by roughly 10 pixels during the observations on both nights, but due to the high quality of the CCD and the use of telescope defocusing this had little impact on the results. We made use of 2 × 2 pixel-binning to achieve a balance between the photometric precision and timing resolution.

We reduced the data using the DEFOT code (Southworth et al. 2014 and references therein), which performs aperture photometry using the APER algorithm and allows for light curves to be obtained relative an optimally weighted ensemble of comparison stars. We used three comparison stars for the night of 2014 September 2 and five for the night of 2014 September 6. The application of bias and flat-field calibrations had a negligible effect on the light curves except for a slight increase in noise, so we elected not to use these calibrations.

2.1.2 CAHA light curves

Light curves of two transits of the target were obtained using the 1.23 m telescope at the Calar Alto Observatory (CAHA), Spain, equipped with the DLR-MKIII CCD camera. This setup yielded a field of view of 21.5 arcmin × 21.5 arcmin at a plate scale of 0.32 arcsec per pixel. We observed through Cousins R, as this filter gave the highest count rate for photons from the target star, and without binning or windowing. The telescope was defocused in order to increase the photometric precision of the observations (Southworth et al. 2009) and using our standard approach for this and other telescopes (Mancini et al. 2014). Although we used autoguiding, we found that the centroid of the target star drifted by roughly 10 pixels during the observations on both nights, but due to the high quality of the CCD and the use of telescope defocusing this had little impact on the results. We made use of 2 × 2 pixel-binning to achieve a balance between the photometric precision and timing resolution.

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2.1.3 TESS light curves

We inferred the TESS light curve of HAT-P-19 b performing a maximum-likelihood estimation of the flux of the target source as well as the fluxes of the neighbouring sources and the background emission. In order to extract the relative flux and uncertainty time-series, we performed PSF photometry on a 11 × 11 pixel region around HAT-P-19 in the full frame images of TESS. We used an empirical estimate of the point response function (PRF)2 of the TESS cameras that was measured while the instrument was operating in space. We obtained a PSF model as a function of sub-pixel shifts by regressing the supersampled PRF to determine the coefficients of polynomial terms up to third order in sub-pixel shifts. For the background emission, we used a spatially flat template justified by the small region of interest.

We retrieved the source positions from TICv8 (Stassun et al. 2019) and corrected them for proper motion. We evaluated the model emission template of each point source by interpolating the PSF model at the respective sub-pixel shift. We assumed that the variance of a pixel is equal to the number of counts in the pixel and neglected inter-pixel covariances. Finally, we performed linear regression to

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Figure 1. T100 light curves of HAT-P-19 b transits in Bessel-R passband.
Table 1. A log of photometric observations analysed within this study. Nightly average of the photometric measurement uncertainties ($\sigma_{ph}$), RMS error of linear fits to out-of-transit levels ($\sigma_{RMS}$), PNR, and the $\beta$-factors quantifying the white and red noise, respectively are given in 4–7 columns.

| Obs. number | Starting date (UT) | Facility | Exp.time (s) | Filter | $\sigma_{ph}$ (mmag) | $\sigma_{RMS}$ (mmag) | PNR (mmag) | $\beta$ |
|-------------|--------------------|----------|-------------|--------|----------------------|-----------------------|------------|--------|
| 1           | 2014-07-28         | T100     | 120         | Bessel-R | 1.03                 | 1.27                  | 0.83       | 0.89   |
| 2           | 2014-08-21         | T100     | 120         | Bessel-R | 0.89                 | 1.06                  | 0.69       | 1.17   |
| 3           | 2015-09-14         | T100     | 120         | Bessel-R | 1.10                 | 1.27                  | 0.83       | 1.11   |
| 4           | 2014-09-02         | CAHA     | 80          | $R_c$   | 0.92                 | 0.97                  | 0.71       | 1.83   |
| 5           | 2014-09-06         | CAHA     | 130         | $R_e$   | 0.72                 | 0.85                  | 0.53       | 1.73   |
| 6           | 2009-10-30         | FLWO     | 150         | Sloan i | 0.90                 | 1.56                  | 0.98       | 1.17   |
| 7           | 2009-12-01         | FLWO     | 90          | Sloan i | 1.13                 | 3.01                  | 2.46       | 1.59   |
| 8           | 2019-10-10         | TESS     | 1800        | TESS    | 0.42                 | 1.38                  | –          | –      |
| 9           | 2019-10-14         | TESS     | 1800        | TESS    | 0.42                 | 1.28                  | –          | –      |
| 10          | 2019-10-22         | TESS     | 1800        | TESS    | 0.42                 | 1.21                  | –          | –      |
| 11          | 2019-10-26         | TESS     | 1800        | TESS    | 0.42                 | 1.11                  | –          | –      |
| 12          | 2019-10-30         | TESS     | 1800        | TESS    | 0.41                 | 0.92                  | –          | –      |

We did not calculate the photon noise rate (PNR), because the integration time for TESS (∼30 min) is comparable to the bin duration (30 min) we made use of in the computation of the PNR, which we determined based on the ingress duration (∼22 min). Although we attempted to calculate the $\beta$ factor, it was based only on 2-point bins covering an hour of integration time giving values between 0.45 and 0.60 contradicting the level of red noise in TESS data. Therefore we do not provide either of these noise statistics in Table 1. The red noise component made it impossible for a single normalization level to be determined for all TESS data set. Hence, we cut the light curve into five segments, each of which has a transit profile at its centre, and we normalized these segments for the analysis to the out-of-transit flux level. Since the transit on 2019 October 18 had only a few points, we ignored it.

2.1.4 Spitzer light curves

We analysed two occultation observations from the Spitzer space telescope in order to better constrain the orbit through $e \cos(\omega)$ parameter. Spitzer observations and light curve extraction were performed by Kammer et al. (2015) using the pixel-level decorrelation method described by Deming et al. (2015). Detailed information (bin size, aperture radius, etc.) about the observations and photometry can be found in table 2 of Kammer et al. (2015). We only analysed two Spitzer occultation light curves at 3.6 μm as presented by Kammer et al. (2015), who obtained two more occultation light curves at 4.5 μm as well. However, they also constrained the prior on eclipse time based on the 3.6 μm observations in their study, because the occultation depths at 4.5 μm turned out to be significantly different than that observed at 3.6 μm. These light curves at 4.5 μm have considerable scatter and the occultation is barely visible in only one of them. Therefore, we only used the two light curves at 3.6 μm to constrain the $e \cos(\omega)$ parameter.

2.2 Light-curve selection for global modelling

In order to select the best light curves in our sample for global analysis we made use of two well-defined metrics, photometric noise rate (PNR) (Fulton et al. 2011) which indicates the white noise and the $\beta$-factor that quantifies the red (or correlated) noise (Winn et al. 2008). PNR is defined by Fulton et al. (2011) as the ratio of the standard deviation of the residuals to the median number exposures per minute including also the time spent for read-out for each observation. The $\beta$-factor, on the other hand, is the ratio of the average residuals in several bins to the standard deviation of the binned residuals. If only the white noise dominates the noise budget, then $\beta = 1$. The ingress/egress time-scale ($\tau$) of HAT-P-19b transits is ∼23 min. Therefore, we grouped our light-curve data points in variable sizes of bins from 13 to 33 min and took the median of $\beta$ factors of those 15 bins in total. Although the $\beta$ factors are comparable only within a group (Winn et al. 2008) of observation runs with the same telescope, two light curves we obtained with 1.23 m telescope of the Calar Alto Astronomical Observatory (CAHA) share similar characteristics with that performed with T100 in that they had been obtained in $R$ filters with similar transmission curves, exposing CCDs for long exposure times in the defocus mode. On the other hand, two light curves from the Fred Lawrence Whipple Observatory’s (FLWO) 1.20 m telescope by Hartman et al. (2011) are in SDSS bands and we have also five TESS transit light curves of the target.

In all of our ground-based observations, the $\beta$ factor, which is a good indicator of the red noise, is between 0.45 and 1.83, showing that white noise is the dominant noise source in all the observations. We had the largest values for this parameter on observations on 2015 October 4 and 28 with T100. The inferior quality of these light curves is also evident from other statistics, more strikingly from the PNR. Therefore, we did not select these light curves in the global modelling. We also had to eliminate the T100 observation on 2015 November 5, although the $\beta$ factor is only 0.85, because the ingress is missing completely, which complicates the modelling as well as decreases the correlated noise artificially as pointed out by Winn et al. (2008). We also had to exclude the light curves acquired on 2016 December 18 transit due to the variations in atmospheric conditions just before the egress. Since these light curves have low-to-moderate quality, either the measured transit depths are significantly different from that observed in all other light curves, when the observing conditions were better or there is an insufficient number of data points at important orbital phases for modelling. Therefore, we decided not to include them in our analysis in obtaining the system parameters. However, it is still possible to determine the mid-transit times with good precision (on the order of a few seconds), hence we used this set of slightly inferior photometric quality observations in the analysis of transit timings. Timestamps in T100 FITS files are updated by the computer clock, which is synchronized with a GPS every few seconds. Therefore the errors in the measurements of mid-
transit times are dominated by photometric precision rather than the timing.

We also computed the nightly average of the photometric measurement uncertainty ($\sigma_{\text{ph}}$) for each night and the root mean square errors of the linear fit to the out-of-transit segments used to detrend the light curves from changing airmass ($\sigma_{\text{RMS}}$). We observed that the transit depths are in one-standard-deviation for all the remaining light curves when these measurement errors are considered. As a result we decided to eliminate five T100 light curves and used the rest in the global modelling although one FLWO light curve (2009 December 1) has considerable scatter, therefore a low PNR. We provide a log of all 12 transit observations in total that we selected for global modelling in Table 1.

### 2.3 Spectroscopic data

High-resolution spectroscopic observations of HAT-P-19 were carried out with High Resolution Echelle Spectrometer (HIRES) attached on the 10 m Keck Telescope between October 2009 and March 2010 and the High-Dispersion Spectrograph (HDS) on the Subaru telescope on Mauna Kea by Hartman et al. (2011) to measure its radial velocity variations due to the transiting object, the planetary nature of which was revealed as a result. For our spectral analysis, we used the only available spectrum in the Keck/HIRES archive, that is free of the absorption lines of $\text{I}_2$ gas due to the iodine cell used to increase radial velocity precision because HDS spectra from Subaru Telescope have narrow wavelength coverage and lower SNR. Average spectral resolution of the Keck/HIRES spectrum is $R \sim 55,000$, which is sufficient to determine the fundamental atmospheric properties of the host star. The spectrum covers a wide wavelength interval between 3360 and 8100 Å. An average SNR of $\sim 130$ was achieved in the 1390 second-exposure on 2010 February 23. This spectrum was also used by Hartman et al. (2011) for the same purpose. We also analysed it to check and confirm the stellar parameters of the host-star and test their findings with different spectroscopic analysis tools, line lists, and stellar atmosphere models.

We corrected the spectrum for both the Doppler shift due to the orbital motion of the star about the common centre of mass with the planet HAT-P-19b and the orbital motion of the Earth about the Sun. We also removed the cosmic rays, and cut the wavelength regions dominated by telluric lines and where the SNR is too low for a thorough analysis. We determined the continuum level from a synthetic spectrum that we used as a visual template for comparison because it is very challenging to determine it from the observational spectrum due to the numerous spectral lines which are blended in most cases. We then normalized the spectrum to this continuum level. We designed this synthetic spectrum in iSpec software package (Blanco-Cuaresma et al. 2014; Blanco-Cuaresma 2019) making use of the lines from Vienna Atomic Line Database (VALD; Piskunov et al. 1995; Kupka et al. 2000; Rybickova et al. 2015) scaled by the ATLAS9 model atmosphere (Castelli & Kurucz 2004) and solar chemical abundances from Asplund et al. (2009), and the Synth code (Kurucz 1993), which is embedded in the iSpec software package. The initial values of spectroscopic parameters were taken from Hartman et al. (2011). In the wavelength regions, where there are significant differences between the line strengths in the observed stellar spectrum and the synthetic spectrum, the extent of such a strong line was accounted for in the determination of the continuum.

The continuum points were marked on the observed stellar spectrum from Hartman et al. (2011). In the wavelength regions, where there are significant differences between the line strengths in the observed stellar spectrum and the synthetic spectrum, the extent of such a strong line was accounted for in the determination of the continuum. The continuum points were marked on the observed stellar spectrum and used as the nodes of cubic spline functions fit to them, and then the spectrum was normalized by dividing the entire spectrum with these spline functions.

**Figure 2.** EWs of 611 Fe-I and 15 Fe-II lines with respect to their abundances for the microturbulent velocity ($\nu_{\text{mic}} = 0.80$ km s$^{-1}$) that minimizes the correlation between the two.

### 3 DATA ANALYSIS AND RESULTS

#### 3.1 Spectroscopy

For stars with effective temperatures lower than 5500 K, techniques based on the measurements of the ratios of equivalent widths (EWs) or depths of certain lines (mostly Fe I and Fe II) (Gray 1994), wings of the $\text{H} \alpha$ line (Fuhrmann 2004), and excitation/ionization balance (Santos, Israelian & Mayor 2004) are not optimum if not totally inadequate in the determination of fundamental parameters due to the excessive line blending (Tsantaki et al. 2013). Synthetic spectrum fitting (Valenti & Fischer 2005) is the preferred technique in this regime of stars; which has been our main method in deriving the fundamental parameters. We looked for the best fitting synthetic spectrum to our observational one with the least-squares minimization method by using the iSpec software package (Blanco-Cuaresma 2019). We tried to fit only the lines of Fe-I and Fe-II having 0.03–1.0 line depths, resulted in $\sim 1300$ lines which we further reduced to 626 in total (611 Fe-I and 15 Fe-II) after having eliminated lines with higher uncertainties than their EWs and beyond the observed ±0.5 dex scatter in Fig. 2 about the mean iron abundance. iSpec also filters out the lines from its analysis, to which it fails to fit a Gaussian. Nevertheless, we still end up with somewhat blended lines that we ignored because it has been shown by Tsantaki et al. (2013) that fundamental parameters ($T_{\text{eff}}, \log g, [\text{Fe/H}]$) derived from an analysis based on only unblended lines to that involving the blended lines do not significantly differ if EWs of iron lines are employed, especially for stars with effective temperatures lower than 5000 K.

For an initial fit to the observed data, $T_{\text{eff}}, \log g, [\text{Fe/H}]$, and line broadening (the sum of rotational and macroturbulent velocities) were adjusted. When a solution is converged, $T_{\text{eff}}, [\text{Fe/H}]$, and broadening amount (in velocity units) were fixed to the values found in the initial fit, while the surface gravity was adjusted to look for a convergence. Since the effect of $\log g$ on the spectra of cool dwarfs is marginal, it will be more adequate to determine its value by using the stellar density derived from the light-curve analysis. In order to keep the surface gravity consistent with the mean density of the star, that we fixed to its value we derived from the light curve analysis for each output surface gravity value, we provided all the fundamental parameters to the equation given by Torres, Andersen & Giménez (2010) as input to derive the stellar mass and radius. We compared the resultant stellar density to the empirical value that we
get from the light-curve analysis. When the difference between the two densities is minimized to 0.001 g cm\(^{-3}\), we accepted the current surface gravity as that of the host star and stopped the iteration, while other parameters were adjusted for a final convergence. The resultant atmospheric model gave the effective temperature, metallicity, and the broadenings (micro/macroturbulence, and the projected rotational velocity), which are summarized in the third column of Table 2.

We measured EWs of 611 Fe-I and 15 Fe-II ‘clean lines’, and determined the microturbulent velocity (\(v_{\text{mic}}\)), for which the EWs become independent of the chemical abundances of the related species, and the scatter in the mean abundances for those is minimized. The value we find for this parameter in this manner is not too different from the microturbulent velocities assumed for HAT-P-19 (\(v_{\text{mic}} = 0.85\) km s\(^{-1}\)) in the literature based on the calibrations (Hartman et al. 2011; Brewer et al. 2016). However, we preferred to fix the microturbulent velocity to the value we have determined from the curve of growth (\(v_{\text{mic}} = 0.80\) km s\(^{-1}\)) given in Fig. 2.

We also attempted at deriving the effective temperature from the ionization balance between Fe-I and Fe-II. However, due to the low temperature of the target, and because there are not many lines of Fe-II in the spectrum with good SNR allowing precise measurements to be made, the effective temperature was underestimated by \(\sim 200\) K. Hence, we rejected the analysis and decided to use the fundamental parameters from the synthetic spectrum fitting. This kind of analysis based on EW measurements gives reliable microturbulent velocities. That is why we used the \(v_{\text{mic}}\) that we determined from this analysis but not the other parameters.

On the other hand, macroturbulent motions and stellar rotation similarly broaden the line profiles, shaping them to be Gaussians. Therefore, the broadening determined by the ISPEC software package from the line profiles is the convolution of both broadening mechanisms. If one of these parameters (e.g. macroturbulence) is fixed to 0, and the other is set free; then the latter (projected rotational velocity (\(v \sin i\)) in our example) is responsible for the entire broadening. That is why we intended to use the typical macroturbulent velocity for our target. Since HAT-P-19 has an effective temperature that is very close to the borderline of the calibration given by Gilmore et al. (2012), we derived the macroturbulent velocity with both calibration equations, one of which is for stars above \(T_{\text{eff}} = 5000\) K (\(v_{\text{mic}} = 3.35\) km s\(^{-1}\)) and the other below that limit (\(v_{\text{mic}} = 2.87\) km s\(^{-1}\)). We found an intermediate value, which is in agreement with the value we found from the weakest Fe-I lines that form deeper in the atmosphere (\(v_{\text{mic}} = 3.00\) km s\(^{-1}\)), after having deconvolved the macroturbulent velocity from the profiles. We then fixed the macroturbulent velocity to this value, and adjusted the projected rotational velocity and found it to be \(v \sin i = 0.88 \pm 1.09\) km s\(^{-1}\). The derived atmospheric parameters from our analysis of the Keck/HIRES spectra of HAT-P-19 are given in Table 2, in comparison with the results from previous works. Our results mostly agree with the previous findings (Hartman et al. 2011; Brewer et al. 2016) although different line lists, techniques,

### Table 2. Atmospheric parameters of HAT-P-19.

| Parameter | H11 | B16 | Initial fit | Final fit |
|-----------|-----|-----|-------------|-----------|
| \(T_{\text{eff}}\) (K) | 4990 ± 130 | 4951 | 4988 ± 54 | 4991 ± 50 |
| \(\log g\) (cgs) | 4.54 ± 0.05 | 4.44 | 4.55 ± 0.07 | 4.53 |
| \([\text{Fe/H}]\) (dex) | 0.23 ± 0.08 | 0.29 | 0.23 ± 0.05 | 0.24 ± 0.05 |
| \(v \sin i\) (km s\(^{-1}\)) | 0.7 ± 0.5 | 1.8 | 1.41 ± 0.87 | 0.88 ± 1.09 |
| \(v_{\text{mic}}\) (km s\(^{-1}\)) | 2.81 | 1.8 | 2.82 ± 0.59 | 3.00 |
| \(v_{\text{mac}}\) (km s\(^{-1}\)) | 0.85 | 0.85 | 0.79 ± 0.15 | 0.80 |

3.2 Global modelling

We used the state-of-the-art, second version of EXOFAST (Eastman 2017, 2019) in order to obtain a global model of the light and radial velocity curves, stellar atmospheric parameters derived from our spectroscopic analysis, and the brightness of the host star in different passbands. We fitted our three best light curves from T100, two light curves from CAHA 1.23 m telescope, two light curves from FLWO 1.20 m observations (Hartman et al. 2011), and all five light curves from TESS that we selected due to their high quality, quantified by the PNR and the \(\beta\) parameters as well as their completeness. We also employed two occultation observations of Spitzer made in 3.6 \(\mu\)m to better constrain the \(e \cos(\omega)\) parameter. We enforced Gaussian priors on the epoch \((R_e = 2456 \ 867.425 \ 742\) in BJD-TDB) based on the mid-transit time of our first observation with T100, and orbital period \((P_{\text{orb}} = 4.008 \ 7826\) d) determined by Hartman et al. (2011). We used the stellar atmospheric parameters we determined as a result of our own analysis of one high resolution, high SNR Keck/HIRES spectrum, as \(T_{\text{eff}} = 4991 \pm 50\) K, \([\text{Fe/H}] = 0.24 \pm 0.05\) dex, and \(v \sin i = 0.88 \pm 1.09\) km s\(^{-1}\) for the mean and standard deviation of Gaussian priors. We did not enforce a prior for the surface gravity since the light curve modelling of exoplanet transits provide better constraints for this parameter, which is not well constrained in spectroscopic analyses of cool star atmospheres. Limb darkening coefficients for the quadratic law were interpolated from the tables provided by

**Figure 3.** Keck/HIRES spectrum of HAT-P-19 (data points) and the best-fitting synthetic spectrum (continuous curve) with iSpec (5290–5320 Å region).

**Figure 4.** Keck/HIRES spectrum of HAT-P-19 (data points) and the best-fitting synthetic spectrum (continuous curve) with iSpec (5160–5200 Å region).
from APASS DR9 catalogue (Henden et al. 2016), while we made the corresponding sky surveys from The Spanish Virtual Observatory catalogue. We took the effective wavelengths of the filters used in 2015), since this value is the same for both point sources in the stellar distance. We collected passband brightness values of HAT-P-19, which is a measure of the star's bolometric flux for HAT-P-19, which is a measure of the star's bolometric flux and independent measurement of the distance, therefore, makes a linear drift in the systemic velocity of HAT-P-19 and fitted was adjusted but limited by the maximum extinction value to that given by the maps from Schlegel, Finkbeiner & Davis (1998).

EXOFAST-v2 can fit a MESA isochrone (MIST) (Paxton et al. 2011, 2013, 2015) and find the position of the host star on the stellar evolutionary tracks based on a trilinear interpolation in Equivalent Evolutionary Phase parameter (EEP), which quantifies the phase at which the host star in its evolutionary history (Eastman 2019), initial stellar mass (\(M_\star\)) and initial metallicity ([Fe/H]0) parameters, which are the mass and surface metallicity value at the zero-age main-sequence (EEP = 202). From the corresponding MESA stellar track, the code derives the effective temperature (\(T_{\text{eff}}\)), stellar radius (\(R_\star\)), and surface metallicity ([Fe/H]) and; compares them to the model values at the current MCMC step. Since we have a well-constrained \(T_{\text{eff}}-\text{[Fe/H]}\) pair from spectroscopy, and \(R_\star\) from SED fitting, the MCMC algorithm constrains the position of the host star on \(T_{\text{eff}}-\log g\) plane by changing \(M_\star\), and [Fe/H]0 to match the values of these parameters by keeping \(M_\star\) consistent with that found from the stellar density (\(\rho_\star\)). The best-fitting MIST evolutionary track for HAT-P-19 from our analysis is given in Fig. 6 with the black continuous curve.

Then the posterior distributions of the global model parameters were calculated by employing likelihood functions utilizing the goodness-of-fit estimators (\(\chi^2\)) from a run of 200 chains of 50,000 iterations. When the variance between the iterations became smaller than the variation inherent to the parameter value, the iterations were aborted by the program, making use of the Gelman–Rubin statistics for the purpose. We provide the median values of all parameters, distributions of which were determined from the global modelling of three T100, two CAHA, two FLWO (Hartman et al. 2011), and five Subaru/HDS observations by Hartman et al. (2011) as they are given, and fitted a velocity offset for each of the TESS light curves, radial velocity values derived from Keck/HIRES and Subaru/HDS observations by Hartman et al. (2011), available passband brightnesses of the target from broad-band photometry, Gaia distance corrected for the systematic shift, and stellar atmospheric parameters determined from our own spectroscopic analysis in Table 4. All the light curves used in the global modelling are given separately in Figs. 7–11.

EXOFAST-v2 models the radial velocity together with the light curves, and then determines the orbital parameters from a global fit. We took the radial velocity values and their errors from Hartman et al. (2011) as they are given, and fitted a velocity offset for each of the HIRES and HDS data sets within every MCMC chain. We also allowed a linear drift in the systemic velocity of HAT-P-19 and fitted

| Passband | \(\lambda_{\text{eff}}\) (Å) | Magnitude |
|----------|-----------------|----------|
| SDSS (Alam et al. 2015) | 3594.9 | 15.589 ± 0.100 |
| SDSS \(u\) | | |
| APASS-DR9 (Henden et al. 2016) | | |
| Johnson \(B\) | 4378.1 | 14.834 ± 0.051 |
| Johnson \(V\) | 5466.1 | 12.853 ± 0.050 |
| SDSS \(g\) | 4640.4 | 13.381 ± 0.176 |
| SDSS \(r\) | 6122.3 | 12.500 ± 0.043 |
| SDSS \(i\) | 7439.5 \(\cos i\) | 12.275 ± 0.174 |
| Pan-STARRS (Chambers et al. 2016) | | |
| zPS | 8657.8 | 12.18 ± 0.05 |
| 2MASS (Skrutskie et al. 2006) | | |
| \(J_{\text{MASS}}\) | 12350.0 | 11.095 ± 0.020 |
| \(H_{\text{MASS}}\) | 16620.0 | 10.644 ± 0.022 |
| \(K_{\text{MASS}}\) | 21590.0 | 10.546 ± 0.019 |
| All WISE (Cutri et al. 2014) | | |
| WISE1 | 33526.0 | 10.495 ± 0.022 |
| WISE2 | 46028.0 | 10.557 ± 0.020 |
| WISE3 | 115608.0 | 10.561 ± 0.091 |
| NOMAD (Zacharias et al. 2005) | | |
| Johnson \(R\) | 6695.6 | 11.99 ± 0.1 |

Figure 5. The passband brightnesses are represented with red data points, the error bars of which in wavelength show the widths of the corresponding filters while those in flux denote the measurement uncertainties. The black continuous curve is the model SED, while blue circles are broadband averages on the model.
3.3 Transit timing variations

We collected light curve data for HAT-P-19 b transits from the literature and the ETD (http://var2.astro.cz/ETD/), where observations of amateur astronomers are also published. We contacted all the observers with quality transit light curves [indicated with a data quality index 2 out of 5 or better (1) as given in ETD]. Based on these communications, we only included light curves from observers, who made use of either a GPS device or a timing server to coordinate their timings throughout their observations, and who provided the time reference for their observations (geocentric JD-UTC, HJD, BJD-UTC, etc.). In the absence of such information, we asked for the raw data from the observer, reduced them ourselves, computed and compared the light curves and the mid-transit times that we obtained with that given in the ETD by the observer. We converted the timings of all observations from the timing reference frames in which they were recorded to Dynamic Barycentric Julian days (BJD-TDB) by using our own scripts that we developed based on the relevant modules and functions of the astropy package (Astropy Collaboration 2013, 2018). As a result, we had a homogenous set of mid-transit times covering almost 10 yr (889 orbital periods) since the discovery of the exoplanet HAT-P-19 b.

We then measured the mid-transit times of all the light curves that we collected (summarized in Table 5) from the literature and ETD, as well as all the light curves we used for modelling, and five additional T100 light curves that we did not use to determine the parameters due to relatively larger photometric errors and interruptions during the observations. We made use of the first EXOFAST version (Eastman 2017) to fit all these light curves separately to measure the mid-transit times in BJD-TDB, forming a homogenous data set from a heterogeneous set of observations. Then we computed the differences between observed and expected times of mid-transits based on the mid-transit time we measured from our best light curve with the minimum scatter [light curve...
We took random, equally probable samples from a parameter space of linear coefficients (slope and y-intercept) and computed the likelihood of a linear fit with these coefficients to the TTV diagram. As a result we obtained the posterior probabilities for the parameters, corresponding to the change in reference mid-transit time ($\Delta T$) and the orbital period ($\Delta P$) within an MCMC run.

We used the Monte Carlo Markov Chain (MCMC) approach to sample the posterior probability distribution of our parameters, marginalizing over uninteresting parameters and allowing for correlated sampling. We used the general purpose MCMC sampler emcee to perform the MCMC runs (Foreman-Mackey et al. 2013).

We iterated over the MCMC along a range of P and $\Delta$P grid and found the best fit for the solar type star HAT-P-19, and its exoplanet HAT-P-19 b using periodicity and TTV analysis. Our best-fit is shown in Fig. 14.

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involving 5000 number of iterations and 500 walkers. We discarded the first 500 steps (the so-called burn-in period) in each of the random walkers until an equilibrium is settled in the search. We refined the ephemeris information with the median values from our MCMC analysis based on all accepted data satisfying the PNR

\[ T_c (\text{BJD-TDB}) = 2456891.47853(3) + 4.008783(3) \times E \]  

and \( \beta \) thresholds from the ETD, literature, and our light curves (equation 1).

We also searched for a periodicity in the frequency space (Fig. 15) and found one frequency with significantly low false alarm probability (FAP \( \sim 0.005 \)) corresponding to a period of 190 d. This frequency arises from seasonal sampling frequency of the observations especially due to the data gaps in data with relatively smaller observational errors, which can be noticed when the data is phased with respect to this frequency. When we remove this frequency from the data, the residuals shows no significant frequencies on the spectrum with the smallest FAP of 0.99 in the remaining frequencies. The linear trend observed in the radial velocities of the target star in Fig. 12 may be indicative of a long period third body, which would be detected in TTVs if there were sufficiently large number of data points spanning a longer baseline because of the large amplitude that the linear increase in RVs implies. However, a probable third body with a different set
Table 5. Mid-transit times measured by us from our light curves, that from the literature and the ETD.

| Mid-transit time (BJD-TDB) | Error (d) | Filter | Observer / reference | RMS of transit fit (norm. flux units) |
|-----------------------------|-----------|--------|-----------------------|--------------------------------------|
| 2455135.63130               | 0.000 17  | Sloan i | Hartman et al. (2011) | 0.0016                               |
| 2455167.70157               | 0.000 16  | Sloan i | Hartman et al. (2011) | 0.0030                               |
| 2455480.38647               | 0.000 90  | Clear  | Lomoz F.              | 0.0053                               |
| 2455496.41966               | 0.001 10  | R_c    | Naves R.              | 0.0035                               |
| 2455524.48256               | 0.000 33  | Clear  | Muler G.              | 0.0020                               |
| 2455528.49220               | 0.000 39  | Clear  | Ruiz J.               | 0.0027                               |
| 2455885.27527               | 0.000 47  | Clear  | Ayiomamitis A.        | 0.0034                               |
| 2455889.27840               | 0.000 25  | Bessel R| Seeliger et al. (2015)| 0.0019                               |
| 2455905.31418               | 0.000 28  | Bessel R| Seeliger et al. (2015)| 0.0021                               |
| 2455913.33317               | 0.000 17  | R_c    | Seeliger et al. (2015)| 0.0012                               |
| 2455921.35082               | 0.000 83  | R_c    | Naves R.              | 0.0027                               |
| 2455921.35282               | 0.001 18  | Clear  | Corfini G.            | 0.0031                               |
| 2456149.85301               | 0.000 53  | Clear  | Shadic S.             | 0.0037                               |
| 2456173.90616               | 0.000 73  | Clear  | Shadic S.             | 0.0047                               |
| 2456270.11566               | 0.000 44  | Clear  | Garlitz J.            | 0.0040                               |
| 2456494.60882               | 0.000 02  | Clear  | Gonzalez J.           | 0.0049                               |
| 2456867.42569               | 0.000 09  | Bessel R| T100 (this study)     | 0.0013                               |
| 2456877.47262               | 0.001 22  | V      | Horta F. G.           | 0.0046                               |
| 2456891.47839               | 0.000 09  | Bessel R| T100 (this study)     | 0.0011                               |
| 2456899.49451               | 0.000 93  | R_c    | Barbieri L.           | 0.0059                               |
| 2456903.50474               | 0.000 11  | R_c    | CAHA (this study)     | 0.0010                               |
| 2456907.51352               | 0.000 11  | R_c    | CAHA (this study)     | 0.0008                               |
| 2456927.55771               | 0.000 41  | R_c    | Gillier C.            | 0.0018                               |
| 2456943.59363               | 0.000 43  | Clear  | Benni P.              | 0.0043                               |
| 2456935.56993               | 0.000 33  | Bessel R| Seeliger et al. (2015)| 0.0025                               |
| 2456947.60245               | 0.000 83  | R_c    | Horta F. G.           | 0.0025                               |
| 2456951.61276               | 0.001 46  | I_c    | Shadic S.             | 0.0099                               |
| 2456975.66639               | 0.001 46  | I_c    | Shadic S.             | 0.0099                               |
| 2457280.33037               | 0.000 09  | Bessel R| T100 (this study)     | 0.0013                               |
| 2457284.33878               | 0.000 42  | Clear  | Ogmen Y.              | 0.0030                               |
| 2457300.37482               | 0.000 26  | Clear  | Maciejewski et al. (2018) | 0.0020                               |
| 2457300.37567               | 0.000 33  | Bessel R| T100 (this study)     | 0.0016                               |
| 2457304.38310               | 0.000 33  | Clear  | Maciejewski et al. (2018) | 0.0019                               |
| 2457316.40888               | 0.000 46  | R_c    | Salisbury M.          | 0.0022                               |
| 2457324.42676               | 0.000 42  | Bessel R| T100 (this study)     | 0.0021                               |
| 2457328.43622               | 0.000 33  | Clear  | Bretton M.            | 0.0016                               |
| 2457332.44554               | 0.000 29  | Bessel R| T100 (this study)     | 0.0010                               |
| 2457336.45472               | 0.000 52  | Clear  | Molina D.             | 0.0027                               |
| 2457340.46237               | 0.000 66  | Clear  | Bretton M.            | 0.0031                               |
| 2457356.50142               | 0.000 90  | Clear  | Bretton M.            | 0.0013                               |
| 2457725.30694               | 0.000 19  | Clear  | Bretton M.            | 0.0020                               |
| 2457725.30828               | 0.000 49  | Clear  | Bretton M.            | 0.0022                               |
| 2457729.31367               | 0.000 49  | Clear  | Bretton M.            | 0.0022                               |
| 2457729.31512               | 0.000 68  | Clear  | Bretton M.            | 0.0023                               |
| 2457733.32484               | 0.000 52  | Clear  | Girardin E.           | 0.0061                               |
| 2457745.35176               | 0.000 37  | Clear  | Bretton M.            | 0.0032                               |
| 2457753.36741               | 0.000 52  | Clear  | Bachschmidt, M.       | 0.0032                               |
| 2457753.36773               | 0.000 52  | Bessel R| T100 (this study)     | 0.0006                               |
| 2458699.43930               | 0.002 70  | Clear  | Jongen, Y.            | 0.0024                               |
| 2458711.46783               | 0.000 50  | Clear  | Raetz, M.             | 0.0024                               |
| 2458723.49466               | 0.000 50  | Clear  | Piirt, P.             | 0.0227                               |
| 2458727.50236               | 0.001 12  | Clear  | Deldem, T.            | 0.0056                               |
| 2458739.52842               | 0.000 72  | Clear  | Jongen, Y.            | 0.0959                               |
| 2458743.53671               | 0.000 99  | Clear  | Grunge, T.            | 0.0038                               |
| 2458747.54616               | 0.000 75  | Clear  | Jongen, Y.            | 0.0019                               |
| 2458755.56291               | 0.000 35  | Clear  | Jongen, Y.            | 0.0326                               |
| 2458767.58860               | 0.000 98  | Clear  | TESS Band             | 0.0004                               |
| 2458771.59776               | 0.000 18  | TESS Band| TESS (this study)     | 0.0013                               |
of potential physical and orbital parameters would escape detection even if it would have a shorter orbital period and the time spanned by the observational TTV data. Therefore, further spectroscopic and transit photometry observations will be needed to probe the nature of the observed TTVs in the system.

4 CONCLUSIONS AND DISCUSSION

4.1 Discussion on the parameters of HAT-P-19 b

Within this work, we analyse a high-resolution Keck/HIRES spectrum to derive the atmospheric parameters of the host star HAT-P-19. We used the results as input parameters for the global modelling based on our high-quality transit light curves, radial velocity measurements by Hartman et al. (2011), and brightness of the star in different passbands that we collected from all-sky survey catalogues to obtain the final parameters of the planet HAT-P-19 b. We made use of the latest (second) version of EXOFAST software package (Eastman 2017, 2019) for the global modelling. Since we have an ultra-precise measurement of the stellar parallax, our SED modelling was based on the most precise distance, which enabled us to derive the stellar radius semi-empirically when the small correction applied by fitting an evolutionary track with a MESA isochrone (MIST) is also accounted for. This resulted in a more precise and accurate value for the planetary radius, which depends less on the theory of stellar evolution. Since the distance value derived from the distance modulus based on absolute and apparent magnitudes of the target in K band ($d = 215 \pm 15$ pc) (Hartman et al. 2011) is not too different from that we derived from the Gaia parallax ($d \sim 200.8^{+3.6}_{-2.4}$ pc), which is part of the reason why our final stellar and planetary parameters are similar to that published in the discovery paper by Hartman et al. (2011), the agreement is within $\pm \sigma$ of the uncertainties end to end.

Based on these parameters and the definitions by Kreidberg (2018), we calculated the scale height to be $H \sim 582$ km and the amplitude of the absorption signal as $422 \pm 18$ ppm at $3.6 \mu$m, and $735 \pm 28$ ppm at $4.5 \mu$m, which makes it a very good candidate for James Webb Space Telescope observations to constrain its atmospheric properties, for which it is already in the prime target list (Moliere et al. 2017).

Stellar parameters of M and K dwarfs are subject to only slight changes during their main sequence lifetimes, which significantly increases the uncertainties on stellar ages derived from isochrone fitting. We found a smaller age value for HAT-P-19 as $5.1^{+4.8}_{-3.5}$ Gyr than found by Hartman et al. (2011) (8.8 $\pm$ 5.2 Gyr). The difference is very subtle. However, Mallonn et al. (2015) determined a gyrochronological age of $5.5^{+1.3}_{-1.1}$ Gyr based on the rotation period of $P_{\text{rot}} = 35.5 \pm 2.5$ d derived from the out-of-transit variation of the host star due to star-spot-induced brightness changes modulated with the rotation. Our estimate for the stellar age is in agreement with their result from their photometric campaign observations within the limits of their uncertainties.

4.2 Discussion on transit timing variations

HAT-P-19 b has some key characteristics, that make it an important target for follow-up observations and further analysis. First of all, the residuals of its radial velocity data from its orbital solution that follow a linear trend may be an indicator of a gravitationally bound, yet unseen object. Such an object would cause TTVs detectable with long-baseline photometry from the ground and space. The alarming observation may be the radial velocity residuals being within a level of $60 \text{ m s}^{-1}$ (Hartman et al. 2011). However, we did not find a statistically significant (FAP $< 0.05$) periodic signal other than the seasonal observation sampling frequency in our Fourier analysis after having removed the linear trend observed in its TTV diagram due to the accumulation of uncertainties in the reference elements ($T_0$ and $P_{\text{rot}}$). Nevertheless, we suggest amateur and professional observers continue their observations of its transits since a longer baseline variation is possible. Because the standard deviation of the scatter of TTV data is $\sim 2.4$ times the mean standard deviation of the observations, there may be a TTV signal with stellar or planetary origin. Some fraction of this signal can be attributed to the activity of the star, which was reported to cause a wavelength-dependent photometric variation of $3.0 - 4.7$ mmag from $I$ to $V$ band, periodically within $35.5 \pm 2.5$ d (Mallonn et al. 2015). We did not detect such a periodicity in our Fourier analysis. However, the observed TTVs might have been affected by either the imperfect measurements of transit timings from asymmetric transit profiles due to stellar variability or imperfect weather conditions at the time, which affect only some of the light curves in our sample obtained with good sensitivity, or quasi-periodic changes in the activity level within a longer cycle. Both aspects of activity phenomenon could also have caused the observed residuals from the Keplerian fit in the radial velocity measurements. However, the amplitude of the residual RV signal ($\sim 60 \text{ m s}^{-1}$) can be found excessive for an activity level observed as out-of-transit photometric variations in the $V$ and $I$ bands. Although there is no correlation between these measurements and the bisection inverse span (BIS) (Hartman et al. 2011), the length of
However, these two fundamental parameters are correlated for the hot-Saturn group of planets. In the mass–radius diagram of 70 at the time of this study according to NASA Exoplanet Archive3 et al. 2015, while the diversity in densities of Neptune-mass planets has already been noticed for planets more massive than Saturn (Bakos et al. 2013), if only the transiting planets are considered, planets with parameters close to but above this limit at around \( P_{\text{orb}} \sim 2.8 \) d, the number of which reached 70 at the time of this study according to NASA Exoplanet Archive3 (Akeson et al. 2013) if only the transiting planets are considered, form the boundary of the desert on the orbital period-mass plane presented here in Fig. 16. This sub-group of hot-Jupiters is often referred to as ‘hot-Saturns’.

They are at the interface of two larger groups in terms of mass, between hot-Jupiters and Neptune-mass planets on the mass–radius diagram (Fig. 17), if \( M_p = 0.1 M_{\text{jup}} \) and \( M_p = 0.4 M_{\text{jup}} \) are taken as the lower and upper mass limits of the former and latter groups, respectively. The lack of correlation between planet mass and radius has already been noticed for planets more massive than Saturn (Bakos et al. 2015), while the diversity in densities of Neptune-mass planets is obvious in Fig. 18, which was also mentioned by Bayliss et al. (2015). However, these two fundamental parameters are correlated for the hot-Saturn group of planets. In the mass–radius diagram of transiting planets less massive than \( 0.8 M_{\text{jup}} \) (Fig. 17), the degree of correlation changes in the transition from Neptune-mass planets after \( M_p \sim 0.1 M_{\text{jup}} \) (twice the mass of the planet Neptune) and that towards hot-Jupiters at \( M_p \sim 0.4 M_{\text{jup}} \). the Pearson correlation coefficient between masses and radii of Neptune-mass planets, defined as the planets between \( 0.025 M_{\text{jup}} \) and \( 0.1 M_{\text{jup}} \), is 0.36, while it is 0.64 for Saturn-mass planets between \( 0.1 M_{\text{jup}} \) and \( 0.4 M_{\text{jup}} \). It then falls to 0.14, when hot-Jupiter group over \( 0.4 M_{\text{jup}} \) (and below \( 0.8 M_{\text{jup}} \)) is considered. We changed both limits by 0.01 \( M_{\text{jup}} \) towards both directions, and watched how the Pearson correlation coefficients and their products change on either side of the limits to be able to find better limits for the group. Although there were a couple of other good candidates, the lack of planets below a density of \( 2.0 \) g cm\(^{-3}\) on the mass–radius diagram in this mass interval, the alignment of hot-Saturns at the right edge of the sub-Jovian desert on the orbital period–mass plane (Fig. 16), and that they form a valley on the mass–density diagram (Fig. 18) also suggest that 0.1 and \( 0.4 M_{\text{jup}} \) form a better pair for the limits of hot-Saturn planets.

Most of the planets in the Saturn-mass group (62 out of 70 in NASA Exoplanet Archive transiting sample) have inflated atmospheres, hence lower mean densities than that of the pure water. There are only \( \rho > 1 \) g cm\(^{-3}\), and two \( \rho > 2 \) g cm\(^{-3}\) cases within these mass limits in our sample of transiting planets. In contrast, there are 10 \( \rho > 2 \) g cm\(^{-3}\) and only four Neptune-mass planets less dense than water between 0.06 and 0.10 \( M_{\text{jup}} \). Nevertheless, it should be noted that planets with larger radii are easier to detect in transit favouring less dense Saturn-mass planets to be found more easily.

Although there is a small denser sub-group, Saturn-mass planets form a valley in between larger-mass hot-Jupiters and smaller-mass Neptunes on the mass–density plane (Fig. 18). Hot-Jupiters forming the upper boundary of the sub-Jovian desert are thought to form beyond the snow-line, then migrate inward after a sudden excitation of eccentricity due to encounters with other planets or stellar bodies, which is followed by tidal circularization stabilizing the orbit at relatively shorter periods (Matskos & Kónigl 2016; Owen & Lai 2018; Szabó & Kálnán 2019). The planetary migration was also suggested to be halted by a sudden loss of density within the accretion

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Figure 16. Orbital period–mass diagram for transiting planets less massive than \( 0.8 M_{\text{jup}} \). Blue circles depict Saturn-mass planets as defined within this study. Grey plus marker is for all others. Red star is HAT-P-19 b, for which the error bars are smaller than the symbol. The boundaries of the sub-Jovian desert as suggested by Mazeh et al. (2016) is overplotted in red-dashed lines. Green dot-dashed line indicates the orbital period, \( P_{\text{orb}} = 2.75 \) d.

Figure 17. Mass–radius diagram (Fig. 16), if only the transiting planets are considered, planets with parameters close to but above this limit at around \( P_{\text{orb}} \sim 2.8 \) d, the number of which reached 70 at the time of this study according to NASA Exoplanet Archive3 (Akeson et al. 2013) if only the transiting planets are considered, form the boundary of the desert on the orbital period-mass plane presented here in Fig. 16. This sub-group of hot-Jupiters is often referred to as ‘hot-Saturns’. They are at the interface of two larger groups in terms of mass, between hot-Jupiters and Neptune-mass planets on the mass–radius diagram (Fig. 17), if \( M_p = 0.1 M_{\text{jup}} \) and \( M_p = 0.4 M_{\text{jup}} \) are taken as the lower and upper mass limits of the former and latter groups, respectively. The lack of correlation between planet mass and radius has already been noticed for planets more massive than Saturn (Bakos et al. 2015), while the diversity in densities of Neptune-mass planets is obvious in Fig. 18, which was also mentioned by Bayliss et al. (2015). However, these two fundamental parameters are correlated for the hot-Saturn group of planets. In the mass–radius diagram of transiting planets less massive than \( 0.8 M_{\text{jup}} \) (Fig. 17), the degree of correlation changes in the transition from Neptune-mass planets after \( M_p \sim 0.1 M_{\text{jup}} \) (twice the mass of the planet Neptune) and that towards hot-Jupiters at \( M_p \sim 0.4 M_{\text{jup}} \). the Pearson correlation coefficient between masses and radii of Neptune-mass planets, defined as the planets between \( 0.025 M_{\text{jup}} \) and \( 0.1 M_{\text{jup}} \), is 0.36, while it is 0.64 for Saturn-mass planets between \( 0.1 M_{\text{jup}} \) and \( 0.4 M_{\text{jup}} \). It then falls to 0.14, when hot-Jupiter group over \( 0.4 M_{\text{jup}} \) (and below \( 0.8 M_{\text{jup}} \)) is considered. We changed both limits by 0.01 \( M_{\text{jup}} \) towards both directions, and watched how the Pearson correlation coefficients and their products change on either side of the limits to be able to find better limits for the group. Although there were a couple of other good candidates, the lack of planets below a density of \( 2.0 \) g cm\(^{-3}\) on the mass–radius diagram in this mass interval, the alignment of hot-Saturns at the right edge of the sub-Jovian desert on the orbital period–mass plane (Fig. 16), and that they form a valley on the mass–density diagram (Fig. 18) also suggest that 0.1 and \( 0.4 M_{\text{jup}} \) form a better pair for the limits of hot-Saturn planets.

Most of the planets in the Saturn-mass group (62 out of 70 in NASA Exoplanet Archive transiting sample) have inflated atmospheres, hence lower mean densities than that of the pure water. There are only \( \rho > 1 \) g cm\(^{-3}\), and two \( \rho > 2 \) g cm\(^{-3}\) cases within these mass limits in our sample of transiting planets. In contrast, there are 10 \( \rho > 2 \) g cm\(^{-3}\) and only four Neptune-mass planets less dense than water between 0.06 and 0.10 \( M_{\text{jup}} \). Nevertheless, it should be noted that planets with larger radii are easier to detect in transit favouring less dense Saturn-mass planets to be found more easily.

Although there is a small denser sub-group, Saturn-mass planets form a valley in between larger-mass hot-Jupiters and smaller-mass Neptunes on the mass–density plane (Fig. 18). Hot-Jupiters forming the upper boundary of the sub-Jovian desert are thought to form beyond the snow-line, then migrate inward after a sudden excitation of eccentricity due to encounters with other planets or stellar bodies, which is followed by tidal circularization stabilizing the orbit at relatively shorter periods (Matskos & Kónigl 2016; Owen & Lai 2018; Szabó & Kálnán 2019). The planetary migration was also suggested to be halted by a sudden loss of density within the accretion

3https://exoplanetarchive.ipac.caltech.edu/
Figure 17. Mass–radius diagram for transiting planets less massive than 0.8 \( M_{\text{Jup}} \). Symbols are the same as in Fig. 16. Vertical dashed lines limit the region where Saturn-mass planets lie. The correlation between mass and radius changes significantly before and after these limits (from \( p = 0.36 \) to \( p = 0.64 \) after 0.1 \( M_{\text{Jup}} \) and down to \( p = 0.14 \) after 0.4 \( M_{\text{Jup}} \). A few density profiles have also been plotted, and their values are given in the legends in g cm\(^{-3}\) units.

Figure 18. Mass–density distribution for transiting planets less massive than 0.8 \( M_{\text{Jup}} \). Symbols are the same as in Fig. 18.

disc, mass-loss due to Roche lobe overflow (Kurokawa & Nakamoto 2014) or tidal thermalization events (Vick, Lai & Anderson 2019). The outer volatile layers of these more massive planets are highly inflated because of the strong irradiation (Mordasini et al. 2015; Lundkvist et al. 2016; Demangeon et al. 2018) whereas less massive planets lose their atmospheres due to photoevaporation and end up at the bottom of the desert (Lopez & Fortney 2014; Owen & Lai 2018; Nielsen et al. 2019). The metallicity dependence of the photoevaporation process was highlighted by Dong et al. (2018) and Petigura et al. (2017). Mass-loss can also occur due to giant planet collisions, which can strip off the atmosphere of a giant planet, leaving its core exposed as a dense Neptunian planet (Mordasini 2018).

The alignment of the shortest-period members of the hot-Saturn group on the period-mass plane forming a wall at the right edge of the sub-Jovian desert between 0.1 and 0.4 \( M_{\text{Jup}} \) also supports this view. Then it can be argued that the planets in the Saturn-mass subgroup seem to be the least massive hot-Jupiters on shortest period orbits \( [P_{\text{orb}} > 2.75 \, \text{d}] \), WASP-49 b (Lendl et al. 2012) is the shortest-period planet of the group with \( P_{\text{orb}} = 2.78 \, \text{d} \) that can survive the photoevaporation process if this mechanism is assumed to be the reason behind the formation of sub-Jovian desert. Therefore, sub-Jovian desert is actually the lack of hot-Saturn planets, as defined between 0.1 and 0.4 \( M_{\text{Jup}} \) mass limits, having orbital periods shorter than \( P_{\text{orb}} \sim 2.75 \, \text{d} \). However, Ionov et al. (2018) claimed that photoevaporation cannot be the only factor for the paucity of sub-Jovians that we here described as hot-Saturns. Most recently, Szabó & Kálmán (2019) have shown the effects of stellar parameters, namely effective temperature, metallicity, and surface gravity in the order of decreasing significance on the boundaries of the desert. The fact that most of the members of the hot-Saturn group have lower densities than both more massive hot-Jupiters and less massive hot-
Neptunes (Fig. 18) also favours explanations predicting a decrease in the average density of the planet such as inflation of the outer layers due to strong stellar irradiation (Giuliot & Showman 2002; Lundkvist et al. 2016; Demangeon et al. 2018), tidal (Jackson, Barnes & Greenberg 2008) and ohmic heating (Batygin & Stevenson 2010; Batygin, Stevenson & Bodenheimer 2011;Laughlin, Crismani & Adams 2011), especially more effective for planets with smaller core masses (Bhatti et al. 2016), and/or interaction with the accretion disc (Kurokawa & Nakamoto 2014).

HAT-P-19 b is a typical example of the more common, less dense group of hot-Saturns closer to the upper mass and lower orbital-period limits, at the middle of the density range spanned by its fellow Saturn-mass planets with bloated atmospheres. It orbits a supersolar metallicity star ([Fe/H] = 0.243), which makes it an interesting example contrasting the cases like HAT-P-12 b (Hartman et al. 2009), WASP-39 b (Faedi et al. 2011), WASP-21 b (Bouchy 2010), and numerous others in this mass regime orbiting metal-poor stars. Planetary radius (Enoch et al. 2011; Faedi et al. 2011), density (Bouchy 2010; Enoch et al. 2011), and core mass (Burrows et al. 2007; Hartman et al. 2011) have been suggested to be correlated with the host-star metallicity for Saturn-mass planets in the past. HAT-P-19 b had been found to be a contradicting example (Hartman et al. 2011) to these correlations, which were, however, based only on a limited sample of Saturn-mass planets known at the time. Equipped with 70 confirmed transiting Saturn-mass planets, we needed to reconstruct these plots (fig. 11 in Hartman et al. 2011 and fig. 6 in Enoch et al. 2011) because they had been based on only a few planets known at the time. We updated these plots with the parameters of 70 transiting planets, for which the masses and radii are known to a few per cent precision. We made use of the NASA Exoplanet Archive by limiting it in the same way as the works suggesting the correlations, computed the core masses by interpolation in the tables given by Fortney et al. (2007) based on the parameters of the planets (masses, radii, and orbital periods) as collected from the NASA Exoplanet archive, and by using the gpc computer code (Giant Planets Cores) developed by Nestor Espinoza,4 which assumes a core composition of 50 per cent rock and 50 per cent ice. We took the iron abundance measurements with respect to the solar-value from the most recent version of the SweetCat Catalogue5 (Santos et al. 2013) as a proxy of the host-star metallicity in order to have a homogenous set. In the end, we were not able to find correlations between either the derived core masses, or the radii with the host-star metallicities. We then further narrowed down our sample with planets having radii between 0.5 and 1.1 R\textsubscript{jup}, which limited our sample size to 46 planets, because Fortney et al. (2007) models give reliable core masses only in this radius regime. We constructed the plot combining the relations given by Hartman et al. (2011) and Enoch et al. (2011), with the mass limits for hot-Saturns as defined in this study, which we present in Fig. 19. Although the core mass and the planet radius are not correlated with the stellar metallicity, density is, as expected to be. It is apparent from the figure that higher the stellar metallicity, higher the core mass, and higher the density with a few exceptions. An interesting feature on the diagram relating the planetary properties to the stellar metallicity is the lack of dense planets orbiting metal-poor stars in the hot-Saturn group with the only exception of Kepler-16 (AB)b (Doyle et al. 2011), which is a circumbinary planet. Nevertheless, we are limited to only 10 planets orbiting metal-poor stars in our sample of hot-Saturns. Potential detections of such planets around metal-poor stars within TESS mission can help us understand if the distribution of planet densities in both sides of the solar iron abundance are similar or not.

On the other hand, we found, for HAT-P-19 b, a negative core mass (M\textsubscript{core} = −3.14 M\textsubscript{Jup}) for 4.5 Gyr-models of Fortney et al. (2007), suggesting a core-free structure basically, and only M\textsubscript{core} = 7.67 M\textsubscript{Jup} for 1 Gyr models, which is less than per cent of the total mass of the planet. Considering the age of the planet, updated within this study as 5.1\textsuperscript{+1.5}_{−1.3} Gyr, it can be suggested that HAT-P-19 b has a small core. However, HAT-P-19 is a supersolar metallicity star with [Fe/H] = 0.24 ± 0.05. Planets orbiting higher metallicity stars have been found to populate the lower boundary of the sub-Jovian desert (Petigura et al. 2017; Dong et al. 2018). Hence, it was argued that the planets that appear closer to the period boundary should have larger cores (Owen & Lai 2018; Szabó & Kalmán 2019), for which HAT-P-19 b and all other inflated hot-Saturns are contradictory examples. In addition, it is also not in line with the argument that the metal content of the proto-planetary environment would lead to a larger core and a smaller radius (Fortney et al. 2007; Spiegel et al. 2014) if it was formed where it is observed now. Therefore, it can be suggested that the in situ formation seems unlikely for this group of planets, although it was found to be plausible for hot-Jupiters and hot-Saturns newly, which should be limited to the rare cases when a sufficiently high-mass core (~15 M\textsubscript{Jup}) triggers runaway gas-accretion (Batygin et al. 2016; Bailey & Batygin 2018). If HAT-P-19 b and other Saturn-mass planets orbiting metal-rich stars had formed in close-vicinity of their host-stars, they could have been expected to have larger-cores and smaller radii (Guillot & Showman 2002; Hartman et al. 2009; Bouchy 2010; Faedi et al. 2011), which is not the case the properties of the observed group suggest.

Owen & Lai (2018) and Giacalone et al. (2017) argued that high-eccentricity migration (Rasio & Ford 1996; Fabrycky & Tremaine 2007) is the most plausible migration mechanism for the planets in the upper mass boundary of the sub-Jovian desert, where the low-density hot-Saturn type planets like HAT-P-19 b lie. Massive planets (M\textsubscript{p} > 0.2 M\textsubscript{Jup}) were suggested to survive closer to their stars although the outer layers of their atmospheres are inflated significantly due to the stellar flux and ohmic heating (Laughlin et al. 2011), they still can keep their H/He dominating bulk composition (Bayliss et al. 2015; Owen & Lai 2018). Nevertheless, relatively smaller projected rotation-rate of HAT-P-19 in the line of sight (v\textsubscript{sin} i = 0.88 km s\textsuperscript{−1}) is in contradiction with the angular momentum transfer expected from the planet to the star during a high-eccentricity migration, leading to a larger rotation rate expected from the age of the star. In such cases (e.g. WASP-33 b: Collier Cameron et al. 2010; WASP-52 b: Mancini et al. 2017) the gyrochronological ages are found to be smaller due to the larger observed rotation rates (Demangeon et al. 2018) than the ages found from isochrones. Therefore, a disc-driven migration (Lin, Bodenheimer & Richardson 1996; Ward 1997; Beaugé & Nesvorný 2013) cannot be ruled out since the projected rotation rate of HAT-P-19 is in agreement with a 5.1 Gyr-old star, whose gyrochronological age was found to be ~5.5\textsuperscript{+1.8}_{−1.3} Gyr by Mallonn et al. (2015). High-eccentricity migration, on the other hand, is the dominating migration mechanism in forming the upper boundary of the sub-Jovian desert as predicted by Owen & Lai (2018), which agrees with the large discrepancies between the gyrochronological ages and that determined from isochrones in majority of the cases (Mxated, Serenelli & Southworth 2016; Demangeon et al. 2018). That said, radii of K-type stars, like HAT-P-19 itself, are underestimated by standard stellar models (Popper 1997). Therefore, the discrepancy between the gyrochronological and isochrone ages is more enhanced for K-stars (Mxated et al. 2016). Magnetic activity and a presence of

4https://github.com/nespinoza/gpc
5https://www.astro.up.pt/resources/sweet-cat/
a close-in giant planet (Hébrard et al. 2013) might be potential causes of this discrepancy other than the migration mechanisms. Since we did not find a discrepancy of the star’s age we found from isochrone fitting and that found by Mallonn et al. (2015) from gyrochronology, and a strong magnetic activity indicator other than a few per cent out-of-transit light curve modulation; HAT-P-19 b might have been migrated to its location via disc-driven migration. In brief, whatever the valid migration mechanism is for HAT-P-19 b, it turned out to be massive enough not to lose its atmosphere due to photoevaporation and/or far enough from its star not to get tidally disrupted.

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DATA AVAILABILITY

The spectroscopic observations underlying this article are available in Keck Observatory Archive at https://www2.keck.hawaii.edu/koa/public/koa.php. TESS and Spitzer light curves were derived from sources in the public domain of Barbara A. Mikulski Archive for Space Telescopes at https://mast.stsci.edu/. Passband brightness data were collected from different sources in literature (see Table 3 for details). FLWO light curves are from Hartman et al. (2011). Some of the light curves to derive mid-transit times were downloaded from Exoplanet Transit Database at http://var2.astro.cz/ETD/. All other light curves appearing for the first time in this article, and the derived mid-transit times will be shared on reasonable request to the corresponding author. Parameters of transiting planets used in the study were gathered from the NASA Exoplanet Archive.

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