Photometry and Transit Modeling of Exoplanet WASP-140b

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Abstract

Eleven transit light curves for the exoplanet WASP-140b were studied with the primary objective to investigate the possibility of transit timing variations (TTVs). Previously unstudied MicroObservatory and Las Cumbres Global Telescope Network photometry were analyzed using Markov Chain Monte Carlo techniques, including new observations collected by this study of a transit in December 2021. No evidence was found for TTVs. We used two transit models coupled with Bayesian optimization to explore the physical parameters of the system. The radius for WASP-140b was estimated to be 1.38 ± 0.17 Jupiter radii, with the planet orbiting its host star in 2.235987 ± 0.000008 days at an inclination of 85.75 ± 0.75 degrees. The derived parameters are in formal agreement with those in the exoplanet discovery paper of 2016, and somewhat larger than a recent independent study based on photometry by the TESS space telescope.

1. Introduction

An exoplanet is, in general, a planet orbiting a star other than our Sun. The first confirmed discoveries of exoplanets were made in the early 1990s, opening up a field that is rapidly expanding with several thousand confirmed exoplanets known today, giving us insight into different planetary systems to our own and introducing challenges to our understanding of how such systems form and evolve.

A variety of techniques are used to discover exoplanets. In this project, we concentrated on the transit methods that have been used to discover the most exoplanets to date—namely monitoring the brightness of the exoplanet system. Exoplanets are generally too close to their host stars to be seen as separate objects. The transit method tracks the brightness of the combined system (exoplanet and host star) with time, looking for changes caused such as when the planet passes in front of its star and blocks some light from reaching the Earth. The method tells us about the size of the planets and the angle at which they orbit about the host star relative to our line of sight.

In this paper we study transits for the exoplanet WASP-140b. This planet was discovered by Hellier et al. (2017), being 2.4 Jupiter masses and orbiting its V = 11.1 K0 host star (R.A. (J2000) 04° 01′ 32.54″, Dec. (J2000) −20° 27′ 03.9″) once in roughly 2.24 days. Hellier et al. note a rotational modulation of the out of transit flux with an ~10.4-day cycle, which they attribute to magnetic activity of the host. They note that the transit is grazing, leading to a higher uncertainty in the estimate radius of the planet (1.44 ± 0.24 Jupiter radii).

We apply the \textit{exotic} model (Zellem et al. 2020) to estimate basic parameters of the system such as time of mid-transit, planetary radius relative to the host star, and orbital radius. We compare and contrast these results with a simple transit model (Mandel and Agol 2002) we implemented with a Bayesian optimizer, as well as with literature results. We were particularly interested in seeing if there were deviations in the times of mid-transits compared to a fixed orbital period. The Transit Timing Variation (TTV) method is based on monitoring such changes in timing of transits. The presence of non-transiting planets (in the same system) can be inferred from TTV measurements. The gravitational interaction of these non-transiting planets will sometimes increase the orbital period of the transiting planet, and at other times decrease the period, depending on their relative positions, and so the mid-transit times will vary from a fixed, regular cycle.

2. Data and initial processing

The bulk of observations are 60-second, unfiltered exposures collected by a 6-inch aperture MicroObservatory (MObs; Sadler et al. 2001) telescope located at Mount Hopkins (latitude 31.675°, longitude −110.952°, 1,268 m altitude above sea level) in Arizona, using a KAF-1403 ME CCD camera with a pixel scale of 5.2″ per pixel and 2 × 2 binning to reduce noise. These data were analyzed using \textit{exotic}, which is a \texttt{PYTHON}-based tool developed by JPL’s “Exowatch” program for reducing exoplanet transit data. This software can run on a variety of operating systems as well as via Google’s online “Collaboratory” tool.

Technical details on \textit{exotic} can be found in Zellem et al. (2020). Priors for Markov Chain Monte Carlo (MCMC) fitting by \textit{exotic} are automatically scraped from the NASA Exoplanet Archive (Akeson et al. 2013), while limb darkening parameters are generated by \textit{exofast} (Eastman et al. 2013). \textit{Exotic} generates 1σ uncertainties based on the resulting posterior distributions.

Only dark images were available for the MObs observations, i.e., no flat field images were collected. The dark frames were collected at the beginning and end of each night of observation. As part of the analysis, \textit{exotic} applied the dark frames to
the science data, and then performed differential aperture photometry. For each transit, the analyst supplied EXOTIC a list of comparison stars. EXOTIC performed a stability assessment of this candidate list, choosing the most stable star as the final comparison star. Relatively poor pointing accuracy of the telescope and drift in tracking throughout a transit could lead to selection of different final comparison stars across the transits. However, typically EXOTIC selected stars 108 or 112 from the AAVSO comparison star sequence for WASP-140. We platesolved science frames for each transit to ensure correct selection of the exoplanet host star, using astrometry.net, together with confirmation using charts prepared using the online AAVSO finding chart tool (VSP).
Figure 2: WASP-140b transit data collected using the LCO. The filters used for the LCO observations are indicated in the appropriate sub-figure captions.

Figure 3. Residuals from linear regression fit of orbits versus mid-transit time for WASP-140b. A linear model was fitted to the residuals, with no statistically significant slope. The grey shaded zone is the 3-σ confidence interval for the regression. The blue line is the mean regression slope, which is not statistically different from zero at the 3-σ level. The error bars for the mid-transit timing estimates overlap with this, and with zero, indicating no statistically significant trends in the residuals. Transits were classified by eye into complete and incomplete transits, to see if data quality might obscure any trends (see Table 3). It does not.

3. Analysis

We analyzed 22 MObs attempts to observe transits of WASP-140b, dating from 12 October 2016 to 24 October 2021. Only 7 resulted in successful measurements of transits (see Figure 1 for charts of representative transits), a success rate of 32%. Clouds or incorrect pointing of the telescope accounted for the failed attempts. Table 1 lists the key output from these fits using EXOTIC, namely the orbital semi-major axis (relative to the stellar radius), the planetary radius, and the time of mid-transit (in BJD). The observations and fitted parameter values from EXOTIC have been uploaded to the AAVSO exoplanet database, under the usercode BTSB.

We also made use of the Las Cumbres Observatory Global Telescope network (LCOGT; Brown et al. 2013), first using archival data of transits and also collecting \( r_p \) photometry on the night of 28 December 2021 using a telescope at the Cerro Tololo Inter-American Observatory. All the analyzed LCOGT data were collected using 0.4-meter telescopes equipped with CCDs. We processed all these data using EXOTIC, following flat fielding, dark subtraction, and bias correction via the LCO BANZAI system.\(^2\) Model fits to the transits are shown in Figure 2 and final parameter estimates are given in Table 1. We did not upload the LCOGT archival data or the model fits based on these to the AAVSO Exoplanet Database, given that we did not collect the data and did not wish to “make claim” to them over the original investigators.
3.1. Orbital period

The ephemeris of Hellier et al. (2017) was used to calculate the number of orbits made by WASP-140b about its host star since their starting epoch. These were then regressed against the mid-transit times given in Table 1 using the “lm” function in R (R Core Team 2021), giving an orbital period of 2.2359870 ± 0.000008 days and an epoch of 2456912.349 ± 0.008. These are in good agreement with the values of Hellier et al. (2017): 2.2359835 ± 0.000008 days for the orbit and 2456912.35105 ± 0.00015 for the epoch. Higher order polynomial fits did not result in additionally statistically significant parameters. Inspection of the residuals (see Figure 3) reveals no apparent variation in period. These results therefore do not indicate any significant transit timing variations (TTVs). As noted above, TTVs would indicate the presence of an additional planet in the WASP-104 system through its gravitational attraction periodically altering the orbital velocity of WASP-140b. This would have led to observed transits (of WASP-140b) being earlier or later than predicted by a linear ephemeris. Maciejewski (2022) also analyzed Transiting Exoplanet Survey Satellite (TESS; Ricker et al. 2015) data for the system searching unsuccessfully for TTVs, concluding that there were none currently detectable and so in agreement with the current study.

3.2. Transit models

While exotic had already fitted the transits, we decided to build from “first principles” a simple transit model and couple this with optimization techniques in order both to make a comparison and to explore, including inclination as a free parameter. This was primarily a student project acting as an introduction to exoplanet research, so building our own model and coupling this with optimization was considered a good learning exercise. exotic adopts its priors from the NASA Exoplanet Archive, hence it adopted the inclination from Hellier et al. (2017) as a fixed parameter. Given the glancing nature of this transit, fixing the inclination has a large effect on the derived parameter estimates. For optimization of our transit model, we used the Markov Chain Monte Carlo (MCMC) technique Hamilton Monte Carlo (HMC). MCMC allows construction of a Markov process such that the stationary distribution is the same as our target distribution, through the generation of a “chain” of random samples from the process. Through a sufficient number of samples, such a chain becomes close enough to the stationary distribution and therefore provides a good approximation to the target distribution. This is known as convergence of the MCMC chain (see Sinharay 2003), and allows exploration of the uncertainty in the parameter estimates—explaining our interest in this technique. We implemented HMC using the rstan implementation of STAN (Carpenter et al. 2017; STAN Development Team 2021) inside the statistical programming language R. Uniform priors were used, reflecting minimum previous knowledge of the parameters.

To build this model we used some key parameters of the exoplanet and its host star:

- a, r_s, and r_p were as defined above, with the radii being in terms of a;
- u = linear limb darkening coefficient (see below for an explanation of this parameter);
- \(i\) = orbital inclination (in degrees). Ninety degrees means that the orbital plane is in the line of sight from the Earth;
- offset = a parameter to adjust the reference point of phase axis;
- U = system brightness, used to adjust the reference point of flux axis. The out-of-transit flux should be approximately unity, i.e., the fluxes are normalized to the mean out-of-transit level.

We first consider that \(d\) is the center-to-center distance between the planet and the star. We can then calculate \(z = d / r_s\), which denotes the normalized separation of the centers (of the exoplanet and its host star) and \(p = r_p / r_s\), which is the ratio of the disk radii. This allows us to model a transit based on the equations in Mandel and Agol’s (2002) paper. These specify that for a uniform source, the ratio of obscured to unobscured flux is \(F^o(p, z) = 1 - \lambda^e(p, z)\), where:

\[
\lambda^e(p, z) = \frac{1}{2} \left( \sin^2 \frac{\pi p}{2} + \sin^2 \frac{\pi z}{2} - \sin \frac{\pi p}{2} \sin \frac{\pi z}{2} \right)
\]

Table 1. Fitted Parameters for WASP-140b from the exotic modelling.

| Date         | Mid-transit | \(a/r_s\) | \(r_p/a\) | \(r_p/r_s\) | Quality |
|--------------|-------------|-----------|-----------|-------------|---------|
| 18 Nov 2018  | 2458441.7633 ± 0.0028 | 7.69 ± 0.30 | 0.130 ± 0.005 | 0.1786 ± 0.0099 | complete |
| 22 Jan 2019  | 2458506.6080 ± 0.0026 | 7.63 ± 0.24 | 0.131 ± 0.004 | 0.179 ± 0.001 | partial |
| 11 Oct 2020  | 2459134.9220 ± 0.0038 | 7.51 ± 0.52 | 0.133 ± 0.002 | 0.154 ± 0.024 | complete |
| 20 Oct 2020  | 2459143.8611 ± 0.0020 | 8.40 ± 0.26 | 0.119 ± 0.004 | 0.178 ± 0.015 | complete |
| 29 Oct 2020  | 2459152.8145 ± 0.0026 | 8.14 ± 0.33 | 0.123 ± 0.005 | 0.176 ± 0.016 | complete |
| 15 Dec 2020  | 2459199.7704 ± 0.0083 | 7.29 ± 0.64 | 0.137 ± 0.030 | partial |
| 02 Jan 2021  | 2459217.6512 ± 0.0023 | 7.70 ± 0.24 | 0.130 ± 0.004 | 0.179 ± 0.012 | complete |
| 04 Oct 2019  | 2458761.5091 ± 0.0004 | 7.631 ± 0.085 | 0.131 ± 0.001 | 0.1684 ± 0.005 | complete |
| 14 Oct 2020  | 2459137.1516 ± 0.0023 | 7.20 ± 0.23 | 0.139 ± 0.008 | 0.1618 ± 0.0085 | complete |
| 24 Oct 2021  | 2459512.8046 ± 0.0033 | 6.56 ± 0.19 | 0.152 ± 0.006 | 0.1678 ± 0.0059 | partial |
| 28 Dec 2021  | 2459577.6402 ± 0.0015 | 6.486 ± 0.035 | 0.154 ± 0.001 | 0.1783 ± 0.0027 | partial |

Note: Mid-transit times are given in Barycentric Julian Dates (Barycentric Dynamical Time), the orbital semi-major axis (a) in terms of the stellar radius (r_s), and the planetary radius (r_p) relative to the stellar radius. exotic outputs a/r_s so a column giving the inverse is given for convenience when comparing with a later model and the literature. Uncertainties are 1σ.

2 See https://github.com/LCOGT/banzai for further information on this data pipeline.
This set of equations describes the flux of planetary systems in the following cases:

1. When the planetary disk does not obscure any portion of the stellar disk. There will be no dimming of the combined light, and so the normalized flux would be 1.

2. When the planetary disk is completely in front of the stellar disk. In the case of a uniformly bright stellar disk, the dimming will scale by the obscured area—which can be calculated by \( r_s^2/r_p^2 \) (or \( p^2 \)).

3. The boundary case when the planetary disk is moving onto or off the stellar disk.

The fourth case in Equation 1 corresponds to the unlikely case of when the planet is larger than (or equal to the same radius as) its host star.

Limb darkening refers to the phenomenon that the brightness of a star appears to decrease from the center to the edge, or limb, of the observed disk. This occurs because a stellar atmosphere increases in temperature with depth. At the center of a stellar disk, an observer “sees” deeper and hotter layers that emit more light compared to at the limbs, where the upper and cooler layers are seen (which produce less light). The “small planet” approximation was used for the transit model, in that the limb darkening value corresponding to the center of the planetary disk projected onto the stellar disk was uniformly applied across the stellar area obscured by the planet. We implemented linear limb darkening for the model to adjust the obscured flux values, i.e., a limb darkening model with only a single term.

Only one of our data sets (LCOGT 04 October 2019) could be reliably fitted with this model, as it had a sufficient signal-to-noise ratio, a well-defined transit, and sufficient observations before and after the transit so that the out-of-transit flux levels were well constrained. Interestingly, we were not able to derive a determinate solution for the 04 October 2019 data set, which by eye appears to be a suitable transit. This would indicate that we have too many free parameters in the fit, a point we will come back to later in the paper. Table 2 presents results of this fitting and some example MObs fits. Clearly we were asking too much of the MObs data when we included inclination and limb darkening as free parameters, as we have physically unreasonable solutions for these data sets. exoFast is a better tool for these high noise data sets. The HCM fit to the LCOGT data is more reasonable.

### 3.3. Comparison with the literature

Hellier et al. (2017) estimated \( r_p/r_s \) as 0.166 ± 0.009, \( \cos i = 0.117 \pm 0.009 \), and \( r_s/a = 0.125 \pm 0.002 \). These figures are in good agreement with the HMC model fit based on the LCOGT data bar for \( \cos i \), with the HCM model corresponding to an inclination of 85.07 ± 0.75 degrees compared to Hellier et al.’s value of 83.3 ± 0.8 degrees. This is within two standard deviations, though.

A comparison with the results from the exoFast model for the same data shows that the orbital radius from the HMC model is substantially larger (at ~9.2 times the stellar radius), as is the planetary radius (exoFast’s 0.131 ± 0.001 \( r_p \) compared to 0.159 ± 0.013). The lack of agreement is puzzling, given that both Hellier et al. and exoFast both integrate the limb darkened fluxes obscured by the planetary disk, suggesting that the small planet approximation is not the primary cause of the difference.

Davoudi et al. (2020) used exofast (Eastman et al. 2013) to model a clear filter 01 January 2017 transit data set of the system, finding the planet’s radius to be 1.190 ± 0.0735 that of Jupiter, which is smaller than Hellier et al. (2017)’s estimate of 1.44 ± 0.03 R_J and this paper’s of 1.38 ± 0.03 R_J (although within the error ranges). No inclination or orbital radius data were supplied by Davoudi et al., so a comparison is not possible.

Alexoudi (2022) applied the emcee sampler to analyze 28 transits from 3 sectors (Sector 4 from 18 October 2018 to 15 November 2018, sector 5 from 15 November 2018 to 11 December 2018, and sector 6 from 20 December 2018 to 10 January 2019). Alexoudi et al. (2021) applied the emcee Bayesian sampler to analyze 28 transits from 3 sectors (Sector 4 from 18 October 2018 to 15 November 2018, sector 5 from 15 November 2018 to 11 December 2018, and sector 6 from 20 December 2018 to 10 January 2019). Alexoudi et al. (2021) applied the emcee sampler to analyze 28 transits from 3 sectors (Sector 4 from 18 October 2018 to 15 November 2018, sector 5 from 15 November 2018 to 11 December 2018, and sector 6 from 20 December 2018 to 10 January 2019).

### Table 2: MCMC results

| Date            | \( r_p/r_s \)   | \( r_s/a \)  | \( u \)    | \( \cos i \) | \( \sigma \) | Observatory |
|-----------------|-----------------|-------------|-----------|-------------|-------------|-------------|
| 04 October 2019 | 0.159 ± 0.013   | 0.109 ± 0.007 | 0.48 ± 0.23 | 0.086 ± 0.013 | 0.0036 ± 0.0001 | LCOGT       |
| 11 October 2020 | 0.35 ± 0.23     | 0.14 ± 0.04  | 0.55 ± 0.30 | 0.16 ± 0.07  | 0.010 ± 0.001  | MObs        |
| 20 October 2020 | 0.32 ± 0.22     | 0.10 ± 0.02  | 0.53 ± 0.28 | 0.11 ± 0.05  | 0.0058 ± 0.0005 | MObs        |
| 02 January 2021 | 0.33 ± 0.20     | 0.11 ± 0.02  | 0.58 ± 0.28 | 0.11 ± 0.05  | 0.0063 ± 0.0005 | MObs        |

Note: Only one of the LCOGT data sets gave a reliable solution. Results of three of the better MObs transits are shown, to demonstrate the lower confidence in the estimated parameter estimates for such data sets (together with an implausibly large “planet”). Uncertainties are 1σ. “Date” is the night of observation.
Figure 4. Example MCMC results for the 4 October 2019 transit of WASP-140b. This represents 4,000 steps in the Markov chain, including the initial steps known as “burn-in.” These steps are excluded from the final results, and are considered a result of starting the optimization in a lower probability set of parameters, leading to movement to the global minimum. Actual runs included 40,000 steps, which unfortunately “overloaded” the plotting software and are therefore not included here. “Ratio” is the ratio of the planetary radius to the stellar one, “orbital” is the ratio of stellar radius to the orbital semi-major axis, “u” is the linear limb darkening co-efficient, “cos i” is the cosine of the inclination, “offset” an adjustment in phase, “L” an adjustment in flux, and “sigma” an estimate of the white noise in the data. The chart provides the distributions of each of these parameters on its diagonal as bar charts, correlations between the variables are given in the upper right, and scatter plots crossing each of the parameters in turn are given in the lower left. Each point in a scatter plot represents a step in the Markov chain. The bold lines are linear regressions to the data, corresponding to the correlation results.

Figure 5. The figure on the left (a) shows the non-normalized Pre-search Data Conditioning Simple Aperture Photometry (PDC_SAP) generated by the TESS team, which has had removed longstanding systematic trends and so provides better data quality than the simple aperture photometry (also available from MAST). Remaining variability is clearly visible, showing these changes are on timescales comparable to that between transits. Hellier et al. (2017) noted residual variation at a 5–9 milli-magnitude amplitude. This range is consistent with the observed remaining variability. The figure on the right (b) shows one of these transits plus the optimal model generated by the HMC code. This transit is the second from the left in the data following the break in the middle of Figure 5a.

sector 31 from 21 October 2020 to 19 November 2020.) of data collected by the TESS space telescope. Alexoudi derived an inclination of $84.30 \pm 0.06$ degrees, $r_s/a = 0.1166 \pm 0.0008$, and $r_p/r_s = 0.1464 \pm 0.0010$. These values are similar to those of the current paper and Hellier et al. (2017), but not within formal uncertainties. Alexoudi noted the differences with Hellier et al., commenting that these could be due to the higher accuracy of the TESS data. As a check, we downloaded two-minute cadence TESS data from MAST (see Figure 5a) and applied the HMC model to a transit (centered on TBJD 2459161.75; see Figure 5b). We found $r_s/a = 0.109 \pm 0.008$, $r_p/r_s = 0.163 \pm 0.016$, and $\cos i = 0.089 \pm 0.016$ ($\approx 84.87^\circ$). The linear limb darkening coefficient was poorly constrained ($0.48 \pm 0.29$). Our model resulted in a larger planetary radius than Alexoudi’s, and very close to those derived from the LCOGT data.

3.4. Recommendations

Problems with the other data sets included the lack of sufficient pre-transit data, which prevented reliable estimates (e.g., the 14 October 2020 data set), while variations in the out-
of-transit flux levels prevented a reliable fit to the 28 December 2021 data set. The increased noise of the MObs data compared to LCOGT data also led to less accurate parameter estimates, especially for ratio of the planetary to stellar radii. It would be interesting to see if additional data processing, such as collection and use of flat fields, would help improve the quality of these data sets.

For transit fittings of this system, we recommend that the pre- and post-transit observations be roughly as long as the actual transit time period, particularly since the host star appears to be active (changing in flux levels) on a short time scale. For instance, the pre-transit flux levels appear to be greater than post-transit for the 28 December 2021 data set, and are a complication for a simple model such as ours.

A further complication is the use of the small planet approximation for a high inclination orbit such as that for WASP-140b; in later projects we intend to apply a graduated limb darkening adjustment to the obscured flux. There is a clear correlation between u with r_p/r_s and r_s/a (see Figure 4), so locking u to a value based on theory could lead to a tighter confidence interval for these two parameters. The parameter u can also be seen to be poorly defined in Figure 4. This suggests that it could be better to set it to a value using theory and include u as a fixed (rather than a free) parameter. See Banks and Budding (1990) for further discussion of the information content of data and the question of over-parameterization. Finally, WASP-140b transits close to the stellar limb where the gradient will be strongest in the limb darkening, further supporting the conclusion above.

The signal-to-noise ratio is clearly important for transit fitting, affecting the accuracy of the MObs fits by our model. Observations with the LCOGT (similar to those presented here) appear to have sufficient “information content” to support the HMC model, providing sufficient data about the shoulders of the eclipse are collected for accurate estimation of the out-of-transit flux level.

4. Summary

This paper presented MCMC modeling of transits of WASP-140b, collected using robotic telescopes of the MObs and LCOGT. These data included a transit in December 2021 collected by the authors. We coded a fitting function based on the equations of Mandel and Agol (2002) and coupled this with an optimizer. Together with the exotica analysis program, two MCMC-based optimization models have been applied to these transits, deriving estimates for the times of mid-transit as well as physical parameters of the system. Inspection of the mid-transit times revealed a linear period with no statistical evidence from the data of transit time variations, i.e., no evidence for the gravitational influence of a non-transit planet on the orbit of WASP-140b.

Results from the two analysis programs (exotica and HMC) were in good agreement, indicating the radius for WASP-140b to be $1.38^{+0.18}_{-0.19}$ Jupiter radii, with the planet orbiting its host star in $2.253987^{+0.000008}_{-0.000008}$ days at an inclination of $85.75^{+0.75}_{-0.75}$ degrees. The derived parameters are in formal agreement with the discovery paper of Hellier et al. (2017), and somewhat larger than a recent independent study based on photometry by the TESS space telescope (Alexoudi 2022).

We were probably too ambitious in our selection of an exoplanet with a high inclination orbit about a host star itself with rapidly changing flux levels (to apply a high parameter model such as the HMC model), but that is part of the learning process. Application of techniques such as Gaussian Processes to model out the host star variations would be a good next step, which would allow combining multiple transits which could be binned together to increase the signal-to-noise ratio and strengthen the information content of the data. We also plan to use our HMC model on more simple systems, such as Kepler-1 (see, e.g., Ng et al. (2021) who applied the Mandel and Agol (2002) models, MCMC, and Gaussian Processes to Kepler space telescope data of Kepler-1b and other systems), which do not have such active host stars and orbits with inclinations closer to 90 degrees, where the model’s deficiencies will be less and the correlation between limb darkening and inclination less confounding. Having made these comments, we still recommend that programming a simple model such as Mandel and Agol (2002) and coupling this with an optimizer is a useful learning exercise, and makes for a useful student project. Our points are rather to choose a more quiet system than the one we did, and to either implement improved handling of limb darkening for highly tilted systems or to choose an exoplanet with an orbit closer to 90° inclination as well as being somewhat smaller relative to its host star (so that the small planet approximation is more valid). If investigation of TTVs is the primary goal of the project, then exotica is an excellent tool for such work.

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