TOI-1518b: A Misaligned Ultra-hot Jupiter with Iron in Its Atmosphere

Samuel H. C. Cabot1, Aaron Bello-Arufe2, João M. Mendonça2, René Tronsgaard2, Ian Wong3,4, George Zhou4, Lars A. Buchhave5, Debra A. Fisher1, Keivan G. Stassun5, Victoria Antoci2,6, David Baker7, Alexander A. Belinski8, Björn Benneke9, Luke G. Bouma10, Jessie L. Christiansen11, Karen A. Collins4, Maria V. Goliguzova, Simone Hayeg12, Jon M. Jenkins13,22, Eric L. N. Jensen14, Richard C. Kidwell Jr15, Didier Laloum16, Bob Massey17, Kim K. McLeod11, David W. Latham11, Edward H. Morgan19, George Ricker19, Boris S. Safonov8, Joshua E. Schlieder20, Sara Seager19,21, Avi Shporer13,22, Jeffrey C. Smith9, Gregory Sordo23, Ivan A. Strakhov8, Guillermo Torres8, Joseph D. Twicken15,23,24, Roland Vanderspek19, Michael Vezie19, and Joshua N. Winn10

1 Yale University, 52 Hillhouse Avenue, New Haven, CT 06511, USA; sam.cabot@yale.edu
2 National Space Institute, Technical University of Denmark, Elektrovej, DK-2800 Kgs. Lyngby, Denmark
3 Department of Earth, Atmospheric and Planetary Sciences, Massachusetts Institute of Technology, Cambridge, MA 02139, USA
4 Center for Astrophysics | Harvard & Smithsonian, 60 Garden Street, Cambridge, MA 02138, USA
5 Vanderbilt University, Department of Physics & Astronomy, 6301 Stevenson Center Lane, Nashville, TN 37235, USA
6 Stellar Astrophysics Centre, Department of Physics and Astronomy, Aarhus University, Ny Munkegade 120, DK-8000 Aarhus C, Denmark
7 Physics Department, Austin College, Sherman, TX 75090, USA
8 Sternberg Astronomical Institute, M.V. Lomonosov Moscow State University, 13, Universitetskij pr., 119234, Moscow, Russia
9 Department of Physics and Institute for Research on Exoplanets, Université de Montréal, Montreux, QC, Canada
10 Department of Astrophysical Sciences, Princeton University, NJ 08544, USA
11 NASA Exoplanet Science Institute–Caltech/IPAC Pasadena, CA 91125, USA
12 University of Saskatchewan, Saskatchewan, Canada
13 NASA Ames Research Center, Moffett Field, CA 94035, USA
14 Dept. of Physics & Astronomy, Swarthmore College, Swarthmore, PA 19081, USA
15 Space Telescope Science Institute, Baltimore, MD, USA
16 Société Astronomique de France, 3 Rue Beethoven, F-75016 Paris, France
17 Villa ’39 Observatory, Landers, CA 92285, USA
18 Department of Astronomy, Wellesley College, Wellesley, MA 02481, USA
19 Department of Physics and Kavli Institute for Astrophysics and Space Research, Massachusetts Institute of Technology, Cambridge, MA 02139, USA
20 NASA Goddard Space Flight Center, 8800 Greenbelt Road, Greenbelt, MD 20771, USA
21 Department of Aeronautics and Astronautics, MIT, 77 Massachusetts Avenue, Cambridge, MA 02139, USA
22 SETI Institute, Mountain View, CA 94043, USA
23 Kotizarovci Observatory, Sarsoni 90, 51216 Viskovo, Croatia

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Abstract

We present the discovery of TOI-1518b—an ultra-hot Jupiter orbiting a bright star (V = 8.95). The transiting planet is confirmed using high-resolution optical transmission spectra from EXPRES. It is inflated, with $R_p = 1.875 \pm 0.053 R_J$, and exhibits several interesting properties, including a misaligned orbit (240.34$^{+0.98}_{-0.86}$ degrees) and nearly grazing transit ($b = 0.9036^{+0.0086}_{-0.0053}$). The planet orbits a fast-rotating F0 host star ($T_{\text{eff}} \approx 7300$ K) in 1.9 days and experiences intense irradiation. Notably, the TESS data show a clear secondary eclipse with a depth of 364 $\pm$ 28 ppm and a significant phase-curve signal, from which we obtain a relative day–night planetary flux difference of roughly 320 ppm and a 5.2$\sigma$ detection of ellipsoidal distortion on the host star. Prompted by recent detections of atomic and ionized species in ultra-hot Jupiter atmospheres, we conduct an atmospheric cross-correlation analysis. We detect neutral iron ($5.2\sigma$), at $K_p = 157^{+68}_{-70}$ km s$^{-1}$ and $V_{\text{eq}} = -16^{+3}_{-2}$, adding another object to the small sample of highly irradiated gas-giant planets with Fe detections in transmission. Detections so far favor particularly in future emission spectroscopy to probe for a thermal inversion.

Unified Astronomy Thesaurus concepts: Exoplanet atmospheres (487); Transit photometry (1709); High resolution spectroscopy (2096)

1. Introduction

Transiting exoplanets—those that pass directly between their host stars and an observer—offer a wealth of information about their systems. The transit itself is detectable through the minuscule fraction of starlight occulted by the planet, which is well within the sensitivity of many current ground- and space-based telescopes. The Kepler (Borucki et al. 2010) and K2 (Howell et al. 2014) missions together yielded thousands of transiting exoplanet candidates, some of which are among the most notable and well-characterized to date. Planets found by surveys such as HATNet (Bakos et al. 2004), KELT (Pepper et al. 2007), and WASP (Pollacco et al. 2006) orbit some of the brightest stars, and have hence been popular targets for atmospheric characterization. Today, the frontier lies with the Transiting Exoplanet Survey Satellite (TESS; Ricker et al. 2014), which is searching for planets transiting bright stars across the entire sky.

The science drivers behind exoplanet transit observations are several-fold. Newly discovered systems improve our baseline understanding of exoplanet populations and distributions...
(Howard et al. 2012; Fressin et al. 2013; Fulton et al. 2017), as well as how their properties may be linked to system architecture and formation scenarios (Lissauer et al. 2011; Fabrycky et al. 2014; Millholland et al. 2017; Weiss et al. 2018). The presence of additional, non-transiting exoplanets can be inferred from transit-timing perturbations (Holman & Murray 2005; Ballard et al. 2011). The host star’s obliquity can be probed by the Rossiter–McLaughlin (RM) effect (Winn et al. 2010; Triaud 2018), which also relates to formation pathways (Dawson & Johnson 2018, and references therein). Finally, transits enable the study of exoplanet atmospheres based on the excess absorption of starlight from high-altitude species (e.g., Seager & Sasselov 1998; Charbonneau et al. 2002; Snellen et al. 2010; Sing et al. 2016). These latter investigations require dedicated spectroscopic follow up.

In this study, we present the confirmation of the TESS transiting planet candidate TOI-1518b, a highly irradiated gas-giant planet possessing iron vapor in its atmosphere. Of the exoplanets discovered by TESS, this is the first high-resolution detection of an atmospheric species. Several TESS candidates have been confirmed as hot Jupiters so far, including HD 202772Ab (Wang et al. 2019), HD 2685b (Jones et al. 2019), TOI-150b (Cañas et al. 2019), HD 271181b (Kossakowski et al. 2019), TOI-172b (Rodriguez et al. 2019), TOI-564b, and TOI-905b (Davis et al. 2019). However, TOI-1518b is unique due to its close-in orbit (1.9 day period) and high level of irradiation from its F-type host star.

The new planet falls within the category of ultra-hot Jupiters (UHJs), which have equilibrium temperatures exceeding 2000 K (Fortney et al. 2008; Parmentier et al. 2018). Many UHJs contain vaporized metals, both neutral and ionized, in their upper atmospheres (e.g., Casasayas-Barris et al. 2018; Hoeijmakers et al. 2018). These metals and the molecules containing them are recognized as strong sources of opacity in the optical and near-ultraviolet regions (Fortney et al. 2008; Lothringer et al. 2020). UHJs often exhibit thermal inversions (Haynes et al. 2015; Evans et al. 2017); however, the exact species responsible for the inversions are debated (Fortney et al. 2008; Lothringer et al. 2018; Gandhi & Madhusudhan 2019). High-resolution spectroscopy has become a common method for detecting important species in UHJ atmospheres, and also serves as a means of probing winds (Louden & Wheatley 2015; Casasayas-Barris et al. 2019) and extended atmospheres (Yan & Henning 2018).

Our paper is organized as follows. In Section 2, we analyze the TESS photometry of TOI-1518. We reproduce the detection of a planet candidate, obtain constraints on its orbital parameters, and report a robust detection of the secondary eclipse and phase-curve modulations. We also present high-resolution spectroscopic observations of the system during transit. This spectroscopic transit is analyzed in Section 3, from which we measure the RM effect and obtain further constraints on the orbit and host star. Section 3 also includes a review of the cross-correlation method for atmospheric characterization, the results of which are presented in Section 4. Finally, we discuss TOI-1518b in the context of previously studied UHJs in Section 5.

2. Observations and System Characterization

This section describes our analysis of the available TESS photometry of TOI-1518, as well as high-resolution optical spectra of the system. We measure the system parameters by simultaneously modeling the transit, secondary eclipse, and full-orbit phase curve. We also fit spectral lines to determine properties of the star. As detailed below, radial velocity (RV) measurements of the system provide some broad constraints. However, the deduced parameters have large uncertainties owing to the rapid rotation speed of the star.

2.1. TESS Photometry

The star TIC 427761355 (also designated as BD+66 1610) was observed by Camera 3 of the TESS instrument during Sectors 17 and 18 (UT 2019 October 7 to November 27). The Quick Look Pipeline (QLP; Huang et al. 2020) detected a likely transit signal in the photometry and flagged the companion as a candidate transiting exoplanet with parameters characteristic of a close-in hot Jupiter. The system was released as a TESS Object of Interest (TOI) with the designation TOI-1518. The full-frame images (FFIs) were processed by the Science Processing Operations Center (SPOC; Jenkins et al. 2016) and made publicly available on the Mikulski Archive for Space Telescopes (MAST).

We obtained the TESS-SPOC HLSP light curves (Caldwell et al. 2020) for TOI-1518 from MAST. The SPOC data include two versions of the photometry at the standard 30 minute cadence: (1) the Simple Aperture Photometry (SAP) light curve, i.e., the raw photometry extracted from the SPOC pipeline-derived photometric aperture (Twicken et al. 2010; Morris et al. 2020), and (2) the Presearch Data Conditioning SAP (PDCSAP) light curve, which has been corrected for common-mode systematics trends shared by other sources on the detector (i.e., co-trending basis vectors, or CBVs), while preserving the key astrophysical signals of interest (Smith et al. 2012; Stumpe et al. 2012, 2014).

The PDCSAP light curve is considerably cleaner than the SAP photometry, and in this paper, we present the analysis of the PDCSAP light curve. For completeness, we carried out an analogous analysis of the SAP light curve. Systematics were modeled using linear combinations of the CBVs, similar to the detrending methodology in the SPOC pipeline. We obtained results that are statistically consistent with the main PDCSAP-derived values to within 1σ. However, there were residual long-term systematics trends even after detrending with the CBVs, which led to a roughly 10% increase in residual scatter from the best-fit light-curve model when compared to the PDCSAP analysis.

Our analysis methodology closely mirrors the techniques utilized in the extensive previous work on TESS phase curves (e.g., Shporer et al. 2019; Wong et al. 2020a, 2020b, 2020c); consult those references for a detailed description of the data processing and light-curve fitting. The full PDCSAP light curve of TOI-1518 is shown in Figure 1. Each TESS Sector consists of two spacecraft orbits, separated by a pause in science observations for data downlink. Momentum dumps are scheduled during each spacecraft orbit to reset the onboard reaction wheels. In Sectors 17 and 18, these occurred twice per spacecraft orbit and are indicated in the plot by vertical blue dashed lines. The momentum dumps induce small discontinuities in the photometry, as well as occasional short-term flux ramps. We therefore divide the light curve into individual segments separated by the momentum dumps and model the remaining systematics within each segment separately. Significant ramps are trimmed prior to the final fit; the trimmed points are shown in Figure 1 in red. The last data segment of Sector 18 is not included in our analysis, due to severe residual

25 https://mast.stsci.edu/
systematics. We also apply a 16-point-wide moving median filter to the light curve, after masking the transits, which is then used to remove $3\sigma$ outliers. The final light curve contains 1845 points, divided among 11 segments.

Visual inspection of Figure 1 reveals coherent flux modulations synchronized to the planet’s orbit, indicative of a phase curve. To examine the harmonic content of the TESS photometry in more detail, we trim the transits and secondary eclipses from the light curve (after correcting for instrumental systematics; see Section 2.2) and generate the Lomb–Scargle periodogram. The result is plotted in Figure 2. We find a very strong signal at the orbital frequency, as well as another significant periodicity at the first harmonic of the orbital period (i.e., two maxima per orbital period).

The phase curve of a star–planet system formally contains contributions from both the planet and the host star (see review in Shporer 2017). Close-in exoplanets are tidally locked, with fixed dayside and nightside hemispheres; as the planet rotates, the viewing geometry changes, resulting in a periodic modulation of the observed atmospheric flux that varies as the cosine of the orbital phase. Massive orbiting companions can also raise a tidal bulge on the host star’s surface, resulting in a periodic flux modulation that comes to maximum at quadrature (i.e., a signal with a leading-order term at the first harmonic of the cosine); this is typically referred to as ellipsoidal distortion. Last, the mutual star–planet gravitational interaction causes Doppler shifting of the star’s spectrum, producing a modulation in the total system flux within the bandpass that can sometimes be detected in visible-light photometry. This so-called Doppler boosting signal has the same phase alignment as the RV signal, i.e., the sine of the orbital phase.

2.2. Full-orbit Phase-curve Model

We fit the full-orbit phase curve with a composite flux model for the planet $\psi_p$ and the star $\psi_*$ (e.g., Wong et al. 2020b, 2020c, 2021):

$$\psi_p(t) = f_p - A_{\text{atm}} \cos(\phi + \delta),$$

(1)

$$\psi_*(t) = 1 - A_{\text{ellip}} \cos(2\phi) + A_{\text{Dopp}} \sin(\phi).$$

(2)

Here, $A_{\text{atm}}$, $A_{\text{ellip}}$, and $A_{\text{Dopp}}$ indicate the semi-amplitudes of the planet’s atmospheric brightness modulation, the star’s ellipsoidal distortion signal, and the Doppler boosting, respectively; the signs are assigned so that the measured amplitudes are positive under normal circumstances. The variables $f_p$ and $\delta$ signify the average relative brightness of the planet across its orbit and the phase shift in the planet’s phase curve, respectively.
We note that the stellar ellipsoidal distortion signal contains additional higher-order terms (e.g., Morris 1985; Shporer 2017). The second-highest amplitude is expected at the second harmonic of the cosine (i.e., \( \cos(2\phi) \)). However, there is no significant power precisely at that harmonic in the Lomb–Scargle periodogram (Figure 2); the weak signal around 1.6 days\(^{-1}\) is centered at a slightly higher frequency than the second harmonic, and is likely attributable to low-level residual systematics in the light curve. Indeed, when fitting for the second-harmonic amplitudes in the light-curve analysis, we do not measure any significantly nonzero amplitudes. Therefore, we do not include any higher-order terms of the ellipsoidal distortion when generating the final set of phase-curve fit results.

The transits and secondary eclipse light curves \((\lambda_e, \lambda_\text{atm})\) are modeled using batman (Kreidberg 2015). The secondary eclipse depth (i.e., total dayside hemisphere flux) is related to the phase-curve parameters via the expression \( D_e = f_p - A_{\text{atm}} \cos(\pi + \delta) \). Likewise, the hemisphere-averaged nightside flux is given by \( D_n = f_p - A_{\text{atm}} \cos(\delta) \). To accurately model the 30 minute exposures during transit and secondary eclipse, we use an oversampling factor of 60, i.e., averaging the flux from 30 s subexposures at each time stamp.

Any remaining systematics trends in each light-curve segment \(k\) are detrended using generalized polynomials in time:

\[
S_{\text{fit}}^{(k)}(t) = \sum_{j=0}^{N} c_j^{(k)} (t - t_0)^j,
\]

where \(t_0\) is the first time stamp of the segment, and \(N\) is the order of the detrending polynomial, which in the final joint fit is set to the order that minimizes the Bayesian information criterion (BIC) for each segment. The optimal polynomial orders for the 11 light-curve segments included in our analysis are 2, 0, 0, 1, 0, 3, 3, 2, 3, 1, and 3. The total astrophysical-plus-systematics light-curve model, normalized to unity, is

\[
F(t) = \frac{\psi_s(t) \lambda_e(t) + \psi_p(t) \lambda_\text{atm}(t)}{1 + f_p} \times S_{\text{fit}}^{(k)}(t).
\]

To obtain an initial set of results from the TESS photometry, we jointly fit all 11 light-curve segments using the affine-invariant Markov chain Monte Carlo (MCMC) sampler emcee (Foreman-Mackey et al. 2013). The free astrophysical parameters in our fit that are unconstrained by any priors include the transit ephemeris (mid-transit time \(T_c\) and orbital period \(P\)), transit-shape parameters (impact parameter \(b\) and scaled semimajor axis \(a/R_*\)), planet–star radius ratio \(R_p/R_*\), and the phase-curve parameters. The predicted Doppler boosting amplitude assuming the RV-derived mass (see Section 2.8) is roughly 2 ppm—significantly smaller than the uncertainties on the phase-curve amplitudes. Therefore, we do not fit the Doppler signal, while allowing \(f_p, A_{\text{atm}}, A_{\text{eclip}}\), and \(\delta\) to vary. We also include a uniform point-uncertainty parameter \(\sigma_k\) for each light-curve segment as a free parameter in order to ensure a reduced \(\chi^2\) value of one and retrieve realistic uncertainties on the astrophysical parameters. The median values of \(\sigma_k\) range from 147 to 190 ppm across the 11 segments.

The low cadence of the photometry and the grazing nature of the planetary transit mean that the stellar limb darkening is not constrained well by the light curve. We employ the standard quadratic limb-darkening law and apply Gaussian priors to each coefficient. The median values are set to the values from Claret (2018), interpolated for the measured stellar parameters (see Section 2.7) of TOI-1518: \(u_1 = 0.28\) and \(u_2 = 0.23\); the width of the Gaussian is generously set to 0.05, which is several times larger than the corresponding range of coefficient values spanned by the stellar parameter uncertainty regions.

From our preliminary fit to the full TESS light curve, we find that the transit is grazing, corresponding to a planet–star radius ratio of \(R_p/R_* = 0.0987 \pm 0.0017\) and well-constrained transit-shape parameters: \(b = 0.9103 \pm 0.0065\) and \(a/R_* = 4.231 \pm 0.064\). We detect the secondary eclipse with a depth of \(\sim 380\) ppm and a significant atmospheric phase-curve modulation with a semiamplitude of roughly 160 ppm. There is a nearly 5\(\sigma\) detection of the ellipsoidal distortion signal from the host star, with a semiamplitude of around 30 ppm.

To probe for deviations from a circular orbit, we also carry out a separate light-curve fit with the orbital eccentricity \(e\) and argument of periastron \(\omega\) as additional free parameters. From the photometry, the orbital eccentricity is mostly constrained by the timing of the secondary eclipse relative to the mid-transit time, and to a much lesser extent, the relative durations of the transit and secondary eclipse. We obtain a tight 2\(\sigma\) upper limit of \(e < 0.01\) (formally, \(e = 0.0031^{+0.0047}_{-0.0025}\)), and the inclusion of \(e\) and \(\omega\) as free parameters is strongly disfavored by the BIC (\(\Delta\text{BIC} = 16\)). The corresponding \(e \cos \omega\) and \(e \sin \omega\) values, which relate to offsets in the secondary eclipse timing and duration, respectively, are \(0.0007^{+0.0016}_{-0.0012}\) and \(-0.0003^{+0.0030}_{-0.0061}\). We therefore conclude that the orbit of TOI-1518b is consistent with being circular.

Due to the relatively short time span contained within each segment, there is a possibility of small correlations between the coefficients in the detrending polynomials and the phase-curve parameters. To examine the effect of our choice of polynomial orders, we experiment with allowing only polynomials up to first order (i.e., no curvature in the systematics model). The results from the corresponding joint fit agree well with the aforementioned values. In particular, the measured secondary eclipse depth, atmospheric brightness modulation amplitude, and stellar ellipsoidal distortion amplitude are statistically consistent at much better than the 1\(\sigma\) level. Therefore, we conclude that the optimized polynomial orders listed above, which include orders as high as 3, do not bias the astrophysical parameters in any significant way.

2.3. Ground-based Light Curves

We acquired ground-based time-series follow-up photometry of TOI-1518 as part of the TESS Follow-up Observing Program (TFOP).\(^{26}\) We used the TESS Transit Finder, which is a customized version of the Tapir software package (Jensen 2013), to schedule our transit observations. The photometric data were extracted using AstroImageJ (Collins et al. 2017).

A full transit was observed from Adams Observatory at the Austin College (Sherman, TX, USA) 0.6 m telescope on UT 2020 January 5 in I-band (\(\lambda_{\text{eff}} = 806\) nm). A nearly full transit portion of a transit was observed from the Whitin Observatory (Wellesley, MA, USA) 0.7 m telescope on UT 2020 January 6 in Sloan g’-band (\(\lambda_{\text{eff}} = 475\) nm). A full transit was observed from the private Observatory of the Mount (Saint-Pierre-du-Mont, France) 0.2 m telescope on UT 2020 January 8 in R-band (\(\lambda_{\text{eff}} = 647\) nm). A full transit was
observed from the Kotizarovci Observatory (Viskovo, Croatia) 0.3 m telescope on UT 2020 January 12 in the Baader R 610 nm longpass band ($R_{\text{long}}$, $\lambda_{\text{cut-on}} = 610$ nm). A full transit was observed from the Villa 39 observatory (Landers, CA, USA) 0.36 m telescope on UT 2020 January 24 in B-band ($\lambda_{\text{eff}} = 442$ nm). An egress was observed from the University of Saskatchewan Observatory (Saskatoon, SK, Canada) 0.3 m telescope on UT 2020 March 23 using an Astrodon Clear with Blue Blocking Filter (CBB; $\lambda_{\text{cut-on}} = 500$ nm). The light-curve data are available at ExoFOP-TESS. The raw ground-based transit light curves are shown in the Appendix.

Table 1
Parameters for the TOI-1518 (TIC 427761355) Planetary System

| QLP/Atlas Parameters | Symbol | Units | Value |
|----------------------|--------|-------|-------|
| Right Ascension (R.A.) | R.A. | ... | 23 29 04.224 |
| Declination (Decl.) | Decl. | ... | +67 02 05.377 |
| V-band Magnitude | V | mag. | 8.952 |

| Transit and Orbital Parameters | | |
| Orbital Period | $P$ | days | 1.902603 ± 0.000011 |
| Mid-transit Time | $T_{\text{c}}$ | BJD$_{\text{TDB}}$ | 2458787.049255 ± 0.000094 |
| Radius Ratio | $R_p/R_*$ | ... | 0.0988 ± 0.0012 |
| Impact Parameter | $b$ | ... | 0.9036 ± 0.0061 |
| Scaled Semimajor Axis | $a/R_*$ | ... | 4.291 ± 0.057 |
| Orbital Eccentricity | $e$ | ... | <0.01 (2σ) |
| Orbital Inclination | $i_p$ | deg. | 77.84$^{+0.23}_{-0.26}$ |

| Phase-curve Parameters | | |
| Average Relative Planetary Flux | $f_p$ | ppm | 204 ± 27 |
| Planetary Phase-curve Amplitude | $A_{\text{am}}$ | ppm | 160.4 ± 6.7 |
| Planetary Phase-curve Offset | $\delta$ | deg. | −0.7 ± 2.2 |
| Stellar Ellipsoidal Distortion Amplitude | $A_{\text{ellp}}$ | ppm | 31.3 ± 6.0 |
| Secondary Eclipse Depth | $D_e$ | ppm | 364 ± 28 |
| Nightside Flux | $D_n$ | ppm | 43 ± 27 |
| Dayside Brightness | $T_d$ | K | 3237 ± 59 |
| Temperature | | | |
| Nightside Brightness | $T_n$ | K | 1700$^{+700}_{-1200}$ |

| Stellar Parameters | | |
| Effective Temperature | $T_{\text{eff}}$ | K | 7300 ± 100 |
| Metallicity | [Fe/H] | ... | −0.1 ± 0.12 |
| Surface Gravity | $\log g$ | ... | 4.1 ± 0.2 |
| Projected Rotational Speed | $v \sin i$ | km s$^{-1}$ | 85.1 ± 6.3 |
| Stellar Mass | $M_*$ | $M_\odot$ | 1.79 ± 0.26 |
| Stellar Radius | $R_*$ | $R_\odot$ | 1.950 ± 0.048 |

| RV Parameters | | |
| RV Semiamplitude | $K_s$ | m s$^{-1}$ | <281 (2σ) |
| Systemic Velocity | $V_{\text{sys}}$ | km s$^{-1}$ | −13.94 ± 0.17 |

| Planetary Parameters | | |

Table 1 (Continued)

| QLP/Atlas Parameters | Symbol | Units | Value |
|----------------------|--------|-------|-------|
| Planet Mass | $M_p$ | $M_\oplus$ | <2.3 (2σ) |
| Planet Radius | $R_p$ | $R_\oplus$ | 1.875 ± 0.053 |
| Orbital Semimajor Axis | $a$ | au | 0.0389 ± 0.0011 |
| Equilibrium Temperature | $T_{\text{eq}}$ | K | 2492 ± 38 |

Note. Relevant observing information is obtained from the TESS Quick Look Pipeline (QLP) and Atlas parameters. The V-band magnitude is obtained from the TESS input catalog (Stassun et al. 2018b). The transit and phase curve parameters are simultaneously obtained from a joint fit of the full-orbit TESS light curve and ground-based full-transit photometric data sets (Section 2.4). Derived parameters (i.e., quantities not directly fit for in the light-curve analysis) are indicated by the superscript †. The stellar parameters are determined by fitting a co-added high-resolution spectrum with a stellar model using Spectroscopy Made Easy and by a model fit to the broadband SED (Section 2.7). The RV parameters are measured from FIES radial velocities (Section 2.8).

The follow-up light curves confirm that the TESS-detected event occurs on target relative to known Gaia stars. We analyze the five transit observations with full event coverage (i.e., excluding the UT 2020 March 23 egress-only light curve) by fitting each time series with batman. The mid-transit time, orbital period, impact parameter, and scaled semimajor axis are constrained by Gaussian priors based on the results of the TESS phase-curve fit (Section 2.2). Similar to our treatment of the TESS-band transit modeling, the limb-darkening coefficients are constrained by priors derived by interpolating the tabulated values in Claret et al. (2013) for the appropriate bandpass to the measured stellar parameters and uniformly applying a Gaussian width of 0.05. In the case of the non-standard $R_{\text{long}}$ filter used for the UT 2020 January 12 observation, we approximate the bandpass with the Cousins $I$-band. The systematics trends in every transit light curve are modeled as a linear combination of the airmass and the width of the target’s point-spread function, along with a constant offset for normalization.

The comparatively low signal-to-noise of the ground-based transit data sets translates to large relative uncertainties on the measured transit depth, exceeding 10% across all five visits. Nevertheless, we obtain $R_p/R_*$ values that are consistent with the measurement from fitting the TESS light curve alone at better than the 2σ level. Similarly, the five ground-based transit depths are mutually consistent to within 2σ, indicating an achromatic transit.

2.4. Joint Photometric Analysis

To leverage the additional time baseline and complementary constraints on transit geometry provided by the follow-up transit light curves, we carry out a joint analysis of the TESS photometry and ground-based observations. The orbital ephemeris, transit-shape, and phase-curve parameters are allowed to freely vary, while the limb-darkening coefficients remain constrained by the previously defined priors. The astrophysical light curve and instrumental systematics are simultaneously modeled for all six data sets in the MCMC analysis.

27 https://exofop.ipac.caltech.edu/tess
The results of our joint fit are listed in Table 1. Figure 3 shows the binned, phase-folded, and systematics-corrected TESS light curve alongside the best-fit phase-curve model. Close-up views of the primary transit and secondary eclipse portions of the light curve are provided in Figure 4. The secondary eclipse and phase-curve modulations are clearly discernible. The detrended ground-based transit light curves are plotted in the Appendix.

The orbital period of 1.902603 ± 0.000011 is measured to ∼1 s precision. We obtain a planet–star radius ratio of $R_p/R_*$ = 0.0988+0.0012−0.0011, which is marginally more precise than the value derived from the TESS light curve alone. Likewise, we find slightly improved values for the impact parameter and scaled semimajor axis: $b = 0.9036±0.0061$, $a/R_*$ = 4.291±0.057. The secondary eclipse depth is measured to more than 12σ significance: $364±28$ ppm. The atmospheric phase-curve modulation has a semi-amplitude of 160.4 ± 6.7 ppm. No significant phase shift in the planet’s phase curve is measured, indicating that the location of maximum brightness on the dayside hemisphere is well-aligned with the substellar point.

The derived nightside flux is $43±27$ ppm. The ellipsoidal distortion signal from the host star is detected at 5.2σ significance, with a semi-amplitude of 31.3 ± 6.0 ppm. All of the phase-curve parameters are statistically identical to the values that we obtain from fitting the TESS light curve independently. The planet’s atmospheric brightness modulation and the star’s ellipsoidal distortion signal are plotted separately in the middle panel of Figure 3.

The full set of marginalized two-parameter posteriors for the fitted astrophysical quantities (excluding the limb-darkening coefficients) is plotted in the Appendix. As expected, due to the grazing nature of the transits and secondary eclipses, there are significant correlations between $b$, $R_p/R_*$, and $f_p$, in addition to the typical degeneracy between $b$ and $a/R_*$.

### 2.5. SPP Speckle Interferometry

TOI-1518 was observed using speckle interferometry on 2020 October 26 with the Speckle Polarimeter (SPP; Safonov et al. 2017) on the 2.5 m telescope at the Sternberg Astronomical Institute of Lomonosov Moscow State University (SAI MSU). The spectral band has a central wavelength of 880 nm and a FWHM of 70 nm. The detector has a pixel scale of 20.6 mas px$^{-1}$, and the angular resolution was 89 mas. The detection limit for faint stellar companions is provided in Figure 5. We did not detect any companion brighter than this limit, e.g., 6.5 mag at 1″.

### 2.6. EXPRES Spectroscopy

EXPRES is an ultra-stable optical spectrograph recently commissioned at the Lowell Discovery Telescope (Levine et al. 2012). It is designed for extreme-precision RV surveys (see Jurgenson et al. 2016; Blackman et al. 2020; Brewere et al. 2020; Petersburg et al. 2020, for details about the instrument specifications and reduction pipeline) and also has the capacity for atmospheric characterization (see, for example, the recent study of ultra-hot Jupiter MASCARA-2b by Hoeijmakers et al. 2020). One transit of TOI-1518b was observed on the night of 2020 August 2, involving 41 ∼300 s exposures. The extracted spectra have a signal-to-noise (S/N) of ∼20–40 for pixels in the continuum. Orders were continuum normalized (Petersburg et al. 2020), and subsequently stitched together to form one-dimensional spectra.

Telluric absorption from O$_2$ and H$_2$O in Earth’s atmosphere was corrected with molecfit (Smette et al. 2015) in the geocentric rest frame using fitting parameters similar to those of Allart et al. (2017). Indeed, telluric modeling with molecfit has become a frequent step in high-resolution optical atmosphere studies (e.g., Casasayas-Barris et al. 2019), and is advantageous over empirical models for resolving some atmospheric spectral features (Langeveld et al. 2021).

### 2.7. Spectroscopic Modeling

Before analyzing the transit, we used Spectroscopy Made Easy (SME 423; Valenti & Piskunov 1996) to infer stellar parameters from the high-resolution spectra. The analysis closely follows that of Brewer et al. (2016), including the choice of fitting parameters and wavelength segments. The model made use of a VALD3 line-list (Ryabchikova et al. 2015), an ATLAS9 atmospheric model (Kurucz 1993; Heiter et al. 2002), and a Gaussian convolution instrument profile with
The difference in vertical scale between the two plots. The difference in out-of-

Figure 5. SPP 5σ contrast curve for TOI-1518 with autocorrelation function 
(ACF) inset. The observations were obtained at λc = 880 nm (FWHM = 70 nm).

Figure 4. Zoomed-in views of the primary transit (left) and secondary eclipse 
(right) of TOI-1518b. The light curves are binned in 3 minute intervals. Note 
the difference in vertical scale between the two plots. The difference in out-of-
octillation baselines primarily reflects the planetary phase-curve modulation. The bottom panels show the corresponding residuals from the best-fit model.

R = 137,000. Microturbulence was fixed at 0.85 km s⁻¹, and 
macroturbulence was scaled to T eff following the parameterization of Brewer et al. (2016). However, the fit was largely insensitive to these parameters because the broadening is completely dominated by stellar rotation. The rotational broadening also prevents a robust fit to abundances of individual species. We opt to solve for a global [M/H] with the assumption of a solar abundance pattern for individual elements.

The true uncertainties on effective temperature (T eff), 
metallicity ([Fe/H]), and rotation speed (v sin i) are difficult to 
gauge (Piskunov & Valenti 2017). The Levenberg–
Marquardt optimization algorithm involves computing a 
curvature matrix at the minimum of the objective function, 
the inverse of which is the covariance matrix. The square root 
of the diagonal elements are the formal uncertainties on the 
parameters, assuming that the dominant source of uncertainty is 
from measurement errors (i.e., Poisson statistics on the spectrum). The actual uncertainty is dominated by systematic 
effects and model errors, as opposed to measurement errors. 
Piskunov & Valenti (2017) describe a method to incorporate 
model errors. It involves measuring the sensitivity of each 
spectral pixel to changes in the parameters and estimating the 
change necessary to reduce the fit residuals to zero. The cumulative distribution function (CDF) of these parameter 
perturbations is then calculated. The central region of each 
CDF gives an estimate of the model error. Piskunov & Valenti (2017) discuss this method in greater detail, and we adopt it for our analysis.

We find that TOI-1518 is a rapidly rotating F0 star with 

v sin i = 85 ± 6 km s⁻¹, which agrees with expectations for 
this spectral type (Nielsen et al. 2013). A fitted [Fe/H] of 

−0.1 ± 0.12 is low for a star hosting a hot Jupiter (Fischer & 
Valenti 2005); only ~4% of planet hosts have [Fe/H] near −0.1. However, the uncertainties on [Fe/H] are large due to the 
widening and blurring of spectral lines (a consequence of the 
rapid rotation), so the star may be more metal-rich than the best-
fit value suggests. The best-fit effective temperature and surface 
gravity are T eff = 6910 ± 445 K and log g = 3.97 ± 0.62, 
respectively. More detailed investigation of the stellar spectrum 
might warrant modeling non-LTE effects in the deepest lines and 
calibrating line positions and log gf values. However, these 
considerations are most important for cooler stars with total 
rotational broadening <10 km s⁻¹ (Brewer et al. 2016), 
and their impact on TOI-1518 is reduced due to the rotation 
speed. Measurements of v sin i and [Fe/H] are listed in 
Table 1. However, we opt to report the better constrained 
measurements of log g and T eff from our spectral energy 
distribution (SED) modeling (see below). Our inferred v sin i 
is used to analyze the RM effect (McLaughlin 1924; Rossiter 
1924) in Section 3.2.

As an independent determination of the stellar parameters, 
we performed an analysis of the broadband SED of the star 
together with the Gaia DR2 parallaxes (adjusted by +0.08 mas 
to account for the systematic offset reported by Stassun & 
Torres 2018), following the procedures described in Stassun & 
Torres (2016), Stassun et al. (2017, 2018a). We took the BT- 
VF magnitudes from Tycho-2, the BVi magnitudes from APASS, 
the JHK_s magnitudes from 2MASS, the W1–W4 magnitudes 
from WISE, the GGP kernels from Gaia, and the NUV magnitude from GALEX. Together, the available photometry spans the full stellar SED over the wavelength range 0.2–22 μm (see Figure 6).

We performed a fit using Kurucz stellar atmosphere models, 
with the free parameters being the effective temperature (T eff), 
metallicity ([Fe/H]), surface gravity (log g), and extinction (A V); the extinction was restricted to the maximum line-of-sight 
value from the dust maps of Schlegel et al. (1998). The resulting fit (Figure 6) has a χ² of 20.3 (with 12 degrees of freedom) and best-fit parameters T eff = 7300 ± 100 K, 

[Fe/H] = 0.0 ± 0.2, log g = 4.1 ± 0.2, and A V = 0.05 ± 0.05. 
The relatively low A V may be surprising considering the low galactic latitude; however, this A V is consistent with the 3D 
dust maps for this system’s position from Green et al. (2019).
Integrating the (unreddened) model SED gives the bolometric flux at Earth, \( F_{\text{bol}} = (6.52 \pm 0.31) \times 10^{-9} \text{ erg s}^{-1} \text{ cm}^{-2} \). Taking \( F_{\text{bol}} \) and \( T_{\text{eff}} \) together with the Gaia DR2 parallax (4.398 \( \pm \) 0.033 mas) gives a stellar radius of \( R_\star = 1.950 \pm 0.048 R_\odot \). In addition, we use \( R_\star \) and \( T_{\text{eff}} \) to obtain an empirical mass estimate of \( M_\star = 1.79 \pm 0.26 M_\odot \), which is consistent with that calculated via the empirical relations of Torres et al. (2010) — \( M_\star = 1.70 \pm 0.12 M_\odot \).

### 2.8. FIES Spectroscopy

Starting on 2020 June 14th and ending on 2021 February 3rd, we monitored TOI-1518 with the Nordic Optical Telescope (NOT; Djupvik & Andersen 2010) using the Fiber-fed Echelle Spectrograph (FIES; Telting et al. 2014). This was done in order to constrain the out-of-transit Doppler motion of the star, although the high rotation rate of the star broadens the spectral lines and makes it difficult to measure. The FIES high-resolution fiber reaches \( R \sim 67,000 \) and covers wavelengths from 3760 Å to 8840 Å with no gaps below 8200 Å. We obtained 22 spectra, which we extract as described in Buchhave et al. (2010) and assign wavelengths using ThAr calibrations taken immediately before and after each exposure. The S/N per resolution element ranges from 49 to 141, measured in the 5500 Å spectral order. We did not include RVs from the EXPRES spectra when constraining the Doppler motion, as this would require an extra instrumental offset parameter for a single night of data.

To extract the radial velocities from the FIES spectra, we perform a least-squares deconvolution (LSD) analysis to derive the spectroscopic broadening profiles from each observation (Donati et al. 1997). We deconvolve each spectrum against a synthetic non-rotating spectral template generated via the ATLAS9 library (Castelli & Kurucz 2003), and fit the resulting line profiles with a kernel incorporating the rotational, instrumental, and macroturbulent components of the line broadening function, similar to the recent analysis of HAT-P-70 by Zhou et al. (2019). The extracted RVs are listed in Table 2.

One point is excluded from the analysis, since it overlaps with the transit. Using the \texttt{radvel} package (Fulton et al. 2018), we model the orbit as circular with no other planets in the system; the stipulation of a circular orbit is in line with the results of our TESS light-curve fit, which indicated a 2\( \sigma \) upper limit on orbital eccentricity of 0.01 (Table 1). We define Gaussian priors for period and time of conjunction (using the values and uncertainties from Table 1), as well as a broad, uniform prior on the RV semiamplitude \( K_\star \). We sample the parameter space with an MCMC analysis using the default \texttt{radvel} setup, and let the software run until it determines that the chains are well-mixed.

The \( K_\star \) posterior distribution peaks near its median at \( 152 \text{ m s}^{-1} \) with a 1\( \sigma \) error of 75 m s\(^{-1} \), i.e., less than 2\( \sigma \) significance. We derive a 95\% upper limit of 281 m s\(^{-1} \).

Adopting the stellar mass determined in Section 2.7 and the orbital inclination determined in Section 3.2, this corresponds to a planetary upper mass limit of 2.3 \( M_{\text{Jup}} \), well within expectations for hot Jupiters.

To determine the systemic velocity, we compute the weighted mean of the measured RVs, \( -14.79 \pm 0.06 \text{ km s}^{-1} \), which must be corrected for an instrumental offset of \( -0.87 \pm 0.16 \text{ km s}^{-1} \), found from standard stars. We arrive at a systemic velocity \( V_{\text{sys}} \) of \( -13.94 \pm 0.17 \text{ km s}^{-1} \). The derived RVs are displayed in Figure 7, with the posterior distribution of \( K_\star \) visualized along with the phase-folded velocities. The observations provide generally good sampling of the orbital phase, and have mean cadence of 11.2 days between adjacent observations; we do not expect the RV signature to arise from sampling artifacts or aliases. However, more data are needed in order to determine if the scatter in the RVs could be caused by one or more additional planets in the system.

### 3. The Spectroscopic Transit

In this section, we describe the methods used to analyze the spectroscopic transit observation from EXPRES. Cross correlation
was performed with the X-COR pipeline, previously used for atmospheric detections in WASP-121b (Ben-Yami et al. 2020; Cabot et al. 2020) and MASCARA-2b (Hoeijmakers et al. 2020). Cross correlation has become a standard approach for exoplanet atmospheric analyses at high resolution (e.g., Snellen et al. 2010; Brogi et al. 2012; Birkby et al. 2013). This method relies on resolving the orbital motion of the planet via its Doppler shift on absorption lines (or more recently, emission lines, as shown by Nugroho et al. 2017 and Pino et al. 2020). While individual lines are generally low-S/N, their contributions may be stacked by cross correlating an atmospheric model with the data. Then, one can analyze the resultant cross-correlation function (CCF). This technique has led to a slew of molecular detections in the near-infrared (NIR), as well as atomic and ion detections in the optical, starting with KELT-9b (Hoeijmakers et al. 2018). Please see Madhusudhan (2019) and Ben-Yami et al. (2020) for more examples of recent atmospheric detections at high resolution. We briefly discuss the relevant methods in the following subsection. We then turn our attention to the RM effect and atmospheric signals present in the CCFs.

### 3.1. Detrending and Cross Correlation

The most prominent features in the time-series spectra of TOI-1518b are absorption lines originating in the stellar photosphere, as well as telluric lines caused by Earth’s atmosphere. As mentioned above, we corrected tellurics by fitting and dividing each spectrum by a molecfit model. The spectra were then linearly interpolated onto a common 0.01 Å wavelength grid in the barycentric rest-frame. We observed a significant narrow sodium absorption component in the original spectra, which is likely due to the interstellar medium. Next, we co-added all out-of-transit spectra into a master $F_{\text{out}}$ and then divided each individual spectrum by $F_{\text{out}}$. Interstellar medium features were removed through division by $F_{\text{out}}$ since we opted to not correct for the RV motion of the star (Casasayas-Barris et al. 2018). Since stellar lines are significantly broadened from rotation, the RV motion has negligible effect on the planet’s transmission spectrum. Remaining broadband variations in the spectra were removed by a high-pass Gaussian filter with a standard deviation of 75 pixels. We restricted our analysis to the region 4000–6800 Å. The S/N falls off at bluer wavelengths, and redder wavelengths suffer from particularly severe telluric absorption. Throughout the analysis, about 1% of the data were masked to avoid particularly low-S/N pixels on the blue edge of the spectrum and within Balmer lines.

Cross correlation was performed between each transmission spectrum and a continuum-subtracted PHOENIX stellar model (Husser et al. 2013). The model parameters were selected from a grid and chosen to be close to the inferred parameters: $T_{\text{eff}} = 7000 \, K$, $\log g = 4.0$, and $[\text{Fe}/\text{H}] = 0.0$. The CCF is essentially a sliding dot product between the observed spectra and the model template. It is defined as a function of time $t$ and velocity $v$:

$$
\text{CCF}(v, t) = \frac{\sum_i f(i|t)m(i|v)w(i)}{\sum_i m(i|v)w(i)} \tag{5}
$$

Here, the observed spectrum $f(i|t)$ corresponds to the flux in pixel $i$ at time $t$. The PHOENIX stellar template, denoted by $m(i|v)$, has been Doppler-shifted by some velocity $v$ and is interpolated onto the observed wavelength grid. The weighting term $w(i)$ is chosen to be the inverse time variance of each pixel, so as to downweight contributions from pixels previously in the cores of stellar or telluric lines. The CCF velocities are a grid spanning $-500$ to $+500 \, \text{km s}^{-1}$ in increments of $2 \, \text{km s}^{-1}$.

### 3.2. Spin–Orbit Misalignment

Although we have isolated the planetary atmospheric transmission spectrum, there are residuals at former locations of stellar lines that arise from the division by $F_{\text{out}}$. While $F_{\text{out}}$ is a good template for the out-of-transit stellar spectrum, the stellar line profiles during transit are distorted because the planet occults part of stellar disk. The projected location of the planet against the stellar disk changes throughout the transit, dependent on its impact parameter $b$ and projected obliquity $\lambda$. The star has a projected rotation speed $v \sin i$, and the flux emitted at each point on the star’s surface is Doppler-shifted by some local velocity. The transit removes part of the integrated stellar flux, and breaks the symmetry between each side of the rotating star. This phenomenon is known as the RM effect. It is observed via the apparent “Doppler shadow” in the CCFs (Collier Cameron et al. 2010a), where a dark trail traces the local velocity of the occulted stellar region.

We model the shadow in a fashion similar to that of Hoeijmakers et al. (2020), and show the steps in Figure 8. First, we fit a double-Gaussian profile (sum of two Gaussians) to the Doppler shadow in each CCF row and record the inner profile’s fitted mean, standard deviation, and amplitude. The inner profile models the core of the Doppler shadow, whereas the
portion of the stellar disk occulted by the planet has a local velocity
\[ v_\perp(t) = x_\perp(t)v \sin i. \] (6)

The orthogonal distance \( x_\perp \) is determined by the position of the planet:
\[ x_\perp(t) = x_p(t) \cos(\lambda) - y_p(t) \sin(\lambda) \] (7)
\[ y_p(t) = -\frac{a}{R_\star} \cos(2\pi\phi) \cos(i_p). \] (9)

Therefore, we can obtain independent constraints on \( a/R_\star, \lambda, \) \( v \sin i, \) and \( i_p \) from the light curve and spectrum fitting (note the distinction between \( i_p \) and stellar inclination \( i \), the latter of which we do not investigate here; however, it also may be probed by considering differential rotation (Cegla et al. 2016)). We run an MCMC routine that samples these parameters and fits the path of the shadow described by the polynomial fit described above. As an initial check, we use uniform priors: \( 2 < a/R_\star < 12, \) \( 0 < \lambda < 2\pi \) and \( 0 < i_p < \pi \). We define Gaussian priors for the rotation speed and global offset:\n\[ v \sin i \sim \mathcal{N}(\mu = 80, \sigma = 50) \text{ km s}^{-1}, \] \[ V_{\text{sys}} \sim \mathcal{N}(\mu = -14.5, \sigma = 2) \text{ km s}^{-1}. \] The results are not strongly dependent on the choice of prior for the global offset, owing mainly to the large rotation speed. The sampler includes 15 walkers with 50,000 steps each. We set the uncertainty on each point equal to the standard deviation of the Gaussian profile. We assume that the difference between each data point and the model is independent and normally distributed. We discard the first 5000 steps and thin the chains by a factor of 40 (approximately the autocorrelation time).

From this initial analysis, we obtain a scaled semimajor axis \( a/R_\star = 2.95^{+0.72}_{-0.72} \). The inclination is in better agreement with Table 1, at \( i_p = 76.1^{+3.3}_{-3.9} \) degrees. We also note a strong correlation between \( \lambda \) and \( i_p \). Next, we rerun the MCMC using photometrically derived priors on \( a/R_\star \) and \( i_p \) in order to establish a tighter constraint on obliquity. The final results of our MCMC analysis, listed in Table 3, show that TOI-1518b is a highly misaligned, retrograde planet, with \( \lambda = 240.34^{+0.93}_{-0.98} \) degrees. Indeed, close-in gas giants around hot stars are commonly misaligned (Winn et al. 2010). Companions with mass \( \gtrsim 3 M_J \) around hot stars are less likely to be found in retrograde orbits (Hébrard et al. 2011; Triaud 2018), but the RV-derived mass of TOI-1518b is below this threshold.

### 3.3. \( K_p - V_{\text{sys}} \) Analysis

Closer inspection of Figure 8 shows a faint, white trail spanning approximately \( \pm 50 \text{ km s}^{-1} \). This feature is a signature of the planet’s atmosphere. Throughout the transit, the planet’s apparent RV changes as it moves toward and then away from the observer, given by
\[ v_p(t) = -K_p \sin(2\pi(t - T_c)/P), \] (10)
where \( K_p \) is the semiamplitude of the planet’s RV. Because the planet orbits close in, the change in velocity is on the order of tens of \( \text{km s}^{-1} \). The CCF at each time \( t \) peaks when the PHOENIX model template is Doppler-shifted by the planet’s velocity, and features in the model line up with features in the

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**Figure 8.** Cross-correlation function between the PHOENIX stellar template and individual transmission spectra. Top panel: CCF annotated with the start of ingress and egress. The Doppler shadow (dark) and atmospheric trail (light) form a “V” shape with a vertex at about \( -30 \text{ km s}^{-1} \). Middle panel: Doppler shadow model as described in the text. Bottom panel: corrected CCF where the Doppler shadow model has been subtracted.
actual transmission spectrum. The result is a trail in the CCFs that traces out a small portion of a sinusoidal curve. The planetary signal may be further enhanced by aligning and co-adding CCF rows, thus stacking the peaks and improving the signal’s S/N. The slope of the CCF trail near transit is completely determined by $K_p$ through Equation (10). It is also offset from 0 by the systemic velocity $V_{\text{sys}}$. It is useful to determine $K_p$ and $V_{\text{sys}}$ by sampling values from a grid and attempting to shift and stack the CCFs for each combination of values (Brogi et al. 2012). The signal is maximized at the correct set of values.

The CCF trail only appears if the cross-correlation template contains features present in the planet’s transmission spectrum. The trail in Figure 8 indicates that the atmosphere contains neutral and/or ionized species present in the PHOENIX spectrum. The absorption line positions and relative strengths are unique to each species. Therefore, we can cross correlate with a model template containing only one species, and then perform the $K_p - V_{\text{sys}}$ analysis to search for an atmospheric signal. If the stacked CCF contains a sufficiently high significance peak, then we confirm the presence of that species in the atmosphere of the planet. Here, we define detection significance (S/N) as the number of standard deviations that the CCF peak lies away from mean of all values, for all combinations of $K_p$ and $V_{\text{sys}}$. Many species of interest are present in the stellar spectrum and have a Doppler shadow in their CCFs. Therefore, after cross correlating with each model template, we scale the shadow model obtained in Section 3.2 by a best-fitting constant value and subtract it from the CCF.

3.4. Transmission Spectrum Model

During a planet transit, a fraction of the stellar light is filtered by the planetary atmosphere. To compute the high-resolution transmission spectra of the planet’s atmosphere, we first need to calculate the opacities of the elements in the atmosphere. In this work, the Fe and Fe$^+$ opacities were computed using the HELIOS-K software (Grimm et al. 2021). Our models for Fe and Fe$^+$ make use of the line-list tables from Kurucz (2018). The lines for both Fe and Fe$^+$ were computed assuming Voigt profiles, 0.032 cm$^{-1}$ spectral resolution, and a fixed line cutoff of 100 cm$^{-1}$. To calculate the transmission spectra, we developed our code based on the simple formalism presented in Gaidos et al. (2017) and Bower et al. (2019). Our model computes the effective tangent height in an atmosphere that was discretized in 200 annuli. The model included some simplifications due to the unknown composition of the atmosphere of TOI-1518b and a weakly constrained planet bulk density: we assumed a surface gravity of log $g = 3$ and an atmosphere in chemical equilibrium. The chemical calculations were done with the open-source code FastChem (Stock et al. 2018), assuming solar metallicities. We include in our model the H$^+$ bound–free and free–free absorption from John (1988). As shown in Kitzmann et al. (2018), the H$^+$ continuum in UHJs is generally between 1 mbar and 10 mbar. Each high-resolution transmission spectrum includes Fe or Fe$^+$ along with the H$^+$ continuum absorption and scattering by H and H$_2$. We generated a grid of high-resolution transmission spectra, assuming isothermal atmospheres ranging from 2000 to 4000 K in steps of 500 K. Following subtraction of the continuum with a sliding maximum filter and convolution with a Gaussian filter to match the EXPRES instrumental resolution, these models serve as cross-correlation templates.

4. Atmospheric Characterization

4.1. Detections

We detect Fe in the atmosphere of TOI-1518b at the 5.2$\sigma$ level. We also report evidence of Fe$^+$ at the 3.4$\sigma$ level. The PHOENIX model, which contains both species in addition to other atoms and ions, yields an enhanced atmospheric detection at 5.9$\sigma$ confidence, while a combined Fe/Fe$^+$ model yields a 5.4$\sigma$ detection. The Doppler shadow correction removes an artifact that otherwise biases detection significances. The $K_p$ and $V_{\text{sys}}$ corresponding to the peak value are consistent across the various templates. For the PHOENIX model, we find $K_p = 163^{+49}_{-30}$ km s$^{-1}$ and $V_{\text{sys}} = -17^{+3}_{-2}$ km s$^{-1}$. For Fe, the values are $K_p = 157^{+68}_{-44}$ km s$^{-1}$ and $V_{\text{sys}} = -16^{+2}_{-2}$ km s$^{-1}$, and for Fe$^+$, they are $K_p = 178^{+61}_{-60}$ km s$^{-1}$ and $V_{\text{sys}} = -18^{+3}_{-3}$ km s$^{-1}$. Uncertainties correspond to the range of $K_p$ and $V_{\text{sys}}$ within a 1$\sigma$ contour around the peak. Because we only sample a small portion of the planet’s orbit, only loose constraints on the semiamplitude $K_p$ are possible. The $V_{\text{sys}}$ found here is offset by about 3 km s$^{-1}$ at the $\sim$1–2$\sigma$ level. This blueshift may indicate winds in the upper atmosphere of the planet (Miller-Ricci Kempton & Rauscher 2012; Casasayas-Barris et al. 2019). Using values in Table 1, we predict a planetary RV semiamplitude of $K_p = 2\pi a \sin i_p / P = 217.4 \pm 6.2$ km s$^{-1}$. This value is higher than the $K_p$ measured from cross-correlation, but still consistent with within the 1$\sigma$ uncertainties. The above detections are depicted in Figure 9.

Equation (5) involves a normalization term in the denominator that allows the CCF to return a physically meaningful quantity (Hoeijmakers et al. 2019). The CCF peak is a weighted average of the depths of individual lines in the transmission spectrum of the planet. In practice, the average depth depends on the weighting used for low-S/N pixels ($w(i)$) and the wavelength range of the cross-correlation; it also does not correspond to the depth of any particular line. However, it provides an order-of-magnitude estimate of typical absorption depths, and hence the altitude of the species in the exoplanet’s atmosphere. We refer to the average absorption depth as $\langle \delta \rangle$, which is equal to the peak value of the stacked CCF over all $K_p$ and $V_{\text{sys}}$ combinations. As shown by Hoeijmakers et al. (2019), Fe lines probe much deeper in the atmosphere than Fe$^+$ lines under chemical equilibrium. While Fe$^+$ lines are stronger in the optical, they are fewer in number; Fe$^+$ absorption is generally
much stronger in the near-ultraviolet (e.g., Sing et al. 2019). We find average absorption depths of \((3.6 \pm 0.8) \times 10^{-4}\) and \((1.5 \pm 0.4) \times 10^{-3}\) for Fe and Fe\(^+\), respectively. (It should be noted that the significance of the Fe\(^+\) signal only indicates evidence of the species, but we can still proceed with using the signal to learn about the planet).

Per Equation (5), the average absorption depth depends on the absolute depths of lines in the data, as well as the relative (but not absolute) depths of lines in the model. The results above are of the same order of magnitude as those for KELT-9b (Hoeijmakers et al. 2019). The height of the atmosphere \((H)\) extends 5–10 scale heights \((H_\text{sc}\), of length hundreds of kilometers for hot Jupiters; Madhusudhan et al. 2014). The excess absorption beyond the transmission spectrum continuum \((R_p/R_\ast)^2\) is approximately \(\delta \approx 2R_pH/R_\ast^2\); in other words, \(H \approx \delta R_\ast^2/2(R_p/R_\ast)^2\). For order-of-magnitude estimates, we use values in Table 1 and assume the base of the atmosphere has a pressure of 0.01 bar (Kitzmann et al. 2018), which is typical for the H\(^-\) continuum of a UHJ. We also take \(H_\text{sc} \approx 880\) km, estimated from the measured \(T_\text{eq}\) and log \(g\), as well as taking the mean molecular weight as \(\mu = 2.3\) for an H\(_2\)-dominated atmosphere; however, \(\mu\) may be affected by H\(_2\) dissociation on the planet’s dayside. While the mass is highly uncertain, we take the posterior median value of 1.4 \(M_J\) in order to estimate log \(g\). The resultant pressures corresponding to the absorption are \(P \approx 6 \times 10^{-4}\) bar for Fe and \(P \approx 2 \times 10^{-7}\) bar for Fe\(^+\). Interestingly, the blueshift is similar between both Fe and Fe\(^+\) signals, suggesting that high-velocity winds might be fairly consistent across various depths in the atmosphere.

The 4000 K Fe model returns the highest-significance detection. The Fe detection significances are 4.2\(\sigma\), 4.7\(\sigma\), and 5.2\(\sigma\) for temperatures of 2000, 3000, and 4000 K, respectively. The cross-correlation signal also decreases significantly during the second half of transit. The Fe detection significance is 4.6\(\sigma\) when using exposures from only the first half of the transit. It drops to 1–2\(\sigma\) if only exposures from the second half are used. This variability could trace differential chemistry between the morning and evening terminators. For example, Ehrenreich et al. (2020) infer a lack of neutral Fe vapor on the dayside terminator of WASP-76b, based on the changing Doppler shift of the cross-correlation peak in each of their exposures.

Hoeijmakers et al. (2020) observe slightly stronger Fe absorption in the second half of a transit of MASCARA-2b, which they suggest could be due to different temperatures or chemistry between terminators. In the case of TOI-1518b,

Figure 9. Atmospheric detections in TOI-1518b and their \(K_p – V_\text{sys}\) maps. The top left corner of each panel indicates the cross-correlation template, and the bottom right corner lists properties of the peak value, including S/N of the detection, the average absorption depth after co-adding CCF rows (\(\langle \delta \rangle\)), and the maximal value of \(K_p\) and \(V_\text{sys}\). In all panels, the Doppler shadow has been corrected per the methods in Section 3.2. For reference, we show results from cross correlation with the PHOENIX spectrum used to model the Doppler shadow, revealing there are species common to both the planet’s atmosphere and star. The subsequent panels show results from cross correlating with templates containing Fe and Fe\(^+\). In each panel, the white dotted lines indicate the \(K_p\) and \(V_\text{sys}\), with the highest signal.
additional transits would help improve our confidence that the observed variability is indeed of physical origin.

4.2. Temperature and Circulation

From the stellar radius, we can use the values of $\frac{R_p}{R_*}$ and $a/R_*$ from our photometric analysis to straightforwardly compute the planet’s radius and orbital semimajor axis: $R_p = 1.875 \pm 0.053 R_\oplus$ and $a = 0.0389 \pm 0.0011$ au. We also utilize the stellar parameters from the SED fit to further characterize the planet’s atmosphere. The relative flux of the planet $D$ in the TESS bandpass, assuming no reflected starlight (i.e., zero geometric albedo), is related to the hemispherically averaged brightness temperature $T_p$ via the following relation (e.g., Shporer 2017):

$$D = \left(\frac{R_p}{R_*}\right)^2 \frac{\int F_\lambda(T_p) \tau(\lambda) \lambda d\lambda}{\int F_\lambda(T_{eff}) \tau(\lambda) \lambda d\lambda}. \quad (11)$$

Here, the stellar and planetary flux spectra are given by $F_\lambda(T_{eff})$ and $F_\lambda(T_p)$, respectively, and $\tau(\lambda)$ is the transmission function of the TESS bandpass. For simplicity, we assume that the planet’s emission spectrum is well-modeled by a blackbody function.

For the stellar spectrum, following the technique described in Wong et al. (2020c), we use PHOENIX stellar models (Husser et al. 2013) and calculate the integrated stellar flux in the denominator of Equation (11) for a grid of stellar parameters in the vicinity of the values derived from the SED fit. We then construct an empirical polynomial function in $T_{eff}$, $[\text{Fe/H}], \log g$ that smoothly interpolates these values. The planet’s brightness temperature can then be fit for using an MCMC routine, with Gaussian priors for $T_{eff}$, $[\text{Fe/H}], \log g$, and $R_p/R_*$ derived from the SED and TESS light-curve fits.

We use the secondary eclipse depth and nightside flux (Table 1) to calculate the corresponding dayside and nightside brightness temperatures of TOI-1518b: $T_d = 3237 \pm 59$ K and $T_n = 1700^{+700}_{-1200}$ K. The extremely high dayside temperature makes TOI-1518b among the hottest exoplanets hitherto discovered, comparable to other UHJs such as WASP-18b (3100 $\pm$ 49 K; Wong et al. 2020b) and WASP-33b (3105 $\pm$ 95 K; von Essen et al. 2020).

We note that any light reflected off the dayside atmosphere (i.e., nonzero geometric albedo) would decrease the contribution of the planet’s thermal emission to the measured secondary eclipse, resulting in a lower inferred dayside brightness temperature. However, at these high temperatures, all known condensate species are expected to be in the vapor phase across the dayside hemisphere, making reflective clouds unlikely (e.g., Helling et al. 2019). This is supported by emission spectrum modeling of other UHJs spanning optical and thermal infrared wavelengths, which break the degeneracy between short-wavelength reflectivity and planetary thermal emission and indicate geometric albedos consistent with zero (e.g., Shporer et al. 2019; Wong et al. 2020b, 2021).

In the broader context of atmospheric circulation, the measured dayside and nightside brightness temperatures reflect the amount of absorbed insolation and the efficiency of day–night heat transport. We can use the simple thermal balance model outlined in Cowan & Agol (2011) to simultaneously constrain the Bond albedo $A_B$ and the recirculation efficiency $\epsilon$. In this parameterization, $\epsilon$ ranges from 0 (no recirculation) to 1 (uniform global temperature). To properly propagate the uncertainties on the stellar and orbital parameters, we use the methodology described in Wong et al. (2020c). Due to the highly uncertain nightside brightness temperature, we retrieve very poor constraints: $A_B < 0.2$ ($\sigma$) and $\epsilon = 0.5 \pm 0.3$. Higher signal-to-noise is required to construct a more precise picture of the atmospheric heat budget. This may be achieved either by including additional visible-wavelength photometry of the system from the TESS Extended Mission or by obtaining full-orbit phase-curve observations at infrared wavelengths, where the planet–star contrast ratio is significantly higher.

5. Discussion and Conclusions

As there have been only a handful of previous detections of iron in UHJs, TOI-1518b adds an important additional data point in our efforts to understand the dynamics and thermal structure in highly irradiated atmospheres. We make a few concluding remarks about the planet below, and then compare it to other recently characterized UHJs.

5.1. TOI-1518b in the Context of Other Iron Detections

Alkali metals (Na and K) have been detected in transmission for numerous hot Jupiters (e.g., Sing et al. 2016). Over the past two years, Fe has also become an increasingly common detected species, albeit mostly in UHJs with $T_{eq} > 2000$ K (Parmentier et al. 2018). Fe traces winds in the upper atmosphere through the systemic velocity offset of the cross-correlation peak, and is also a potential non-oxide contributor to thermal inversions (Lothringer et al. 2018). In the literature, Fe has been detected in transmission in the following exoplanets: KELT-9b (Hoeijmakers et al. 2018, 2019), WASP-121b (Cabot et al. 2019), MASCARA-2b (Hoeijmakers et al. 2020; Stangret et al. 2020), WASP-76b (Ehrenreich et al. 2020), and TOI-1518b (this study). Fe has been detected in emission in KELT-9b (Pino et al. 2020), WASP-189b (Lendl et al. 2020), and WASP-33b (Yan et al. 2020). These targets are listed in Table 4.

Interestingly, Cauley et al. (2020) do not detect Fe in transmission in WASP-189b, despite it being one of the brightest and hottest systems and the fact that Fe is detected in emission (however, the observations were made under poor weather conditions). We note that, although Ca$^+$ was found in transmission in WASP-33b (Yan et al. 2019), and Fe in emission (Yan et al. 2020), there has been no claim of Fe in transmission. Fe may be especially difficult to detect in WASP-33b due to stellar pulsations. We acknowledge a few additional recent studies, including the nondetection of Fe in WASP-19b (Sedaghati et al. 2021) which is listed in Table 4 (however, this target is considerably fainter than the others, at $V = 12.3$), a recent transmission spectroscopy study of HD149026b (Ishizuka et al. 2021; however, the Fe signal was only at 2.8$\sigma$), and a nondetection in TOI-1431b (which orbits a relatively bright $V = 8.0$ star; this target is listed in Table 4).

While the statistical sample is small, Fe detections seem to favor particularly inflated UHJs, potentially with a cutoff around 1.7–1.8 $R_\odot$. One explanation is that Fe detections require particularly large atmospheric scale heights in order for the atoms to imprint sufficiently deep absorption lines on top of the continuum of the transmission spectrum. However, the surface gravity, which is inversely proportional to scale height, does not show a discernible relationship to Fe detections. For example, Fe was detected in transmission in KELT-9b, whose...
large mass yields a similar log \( g \) as WASP-189b. The log \( g \) of TOI-1518b is less than 3.229 at 95% confidence. There are a few bright targets with \( R_p < 1.7 R_J \) that are without detailed, cross-correlation atmospheric analyses, and do not have reported detections of Fe in transmission: MASCARA-1b (Talens et al. 2017), KELT-7b (Bryla et al. 2015), and KELT-17b (Zhou et al. 2016). As more gas giants are detected and characterized, it will be interesting to see if such a trend between Fe detection and planetary radius continues to hold.

### 5.2. Photometric Mass Measurement and Caveats

In our analysis of the TESS photometry, we obtain a strong detection of the ellipsoidal distortion component of the phase-curve variability. This signal is driven by the tidal response of the stellar surface to the mutual star–planet gravitational interaction, which in turn depends on the mass ratio between the two components. It follows that the measured amplitude of the ellipsoidal distortion signal can be used to obtain an independent estimate of the planet’s mass.

The ellipsoidal distortion of the star is formally modeled as a series of cosine terms, with the semiamplitude of the leading term (at the first harmonic of the orbital phase) related to fundamental parameters of the system via the following expression (e.g., Morris 1985; Shporer 2017):

\[
A_{\text{ellip}} = \alpha_{\text{ellip}} \frac{M_p}{M_*} \left( \frac{R_*}{a} \right)^3 \sin^2 \iota_p.
\]  

(12)

Here, the pre-factor \( \alpha_{\text{ellip}} \) is a function of the linear limb-darkening and gravity-darkening coefficients \( u \) and \( g \) for the host star:

\[
\alpha_{\text{ellip}} = \frac{3}{20} \frac{(u + 15)(g + 1)}{3 - u}.
\]  

(13)

Similar to our treatment of the quadratic limb-darkening coefficients in the TESS phase-curve analysis (Section 2.2), we construct Gaussian priors for \( u \) and \( g \) using values interpolated from the coefficients listed in Claret (2017): \( u = 0.41 \pm 0.05 \) and \( g = 0.12 \pm 0.05 \). We then use Equations (12) and (13) to construct the posterior for \( M_p \) through Monte Carlo sampling of the distribution of values for \( A_{\text{ellip}}, a/R_*, \iota_p, M_*, u, \) and \( g \). We obtain a photometric mass estimate of \( M_p = 4.8^{+1.3}_{-1.1} M_J \). This value is significantly (2.3σ) larger than the RV-derived mass upper limit of 2.3 \( M_J \).

This discrepancy between the phase-curve-derived and RV-derived masses may be attributable to oversimplifications in the stellar tidal response formalism. Gomel et al. (2021) found a discrepancy of up to 30% between the amplitudes of the ellipsoidal distortion derived from the analytic expressions of Morris (1985) and those derived numerically. More fundamentally, the classical theory of stellar ellipsoidal distortion from which Equations (12) and (13) are derived makes several key assumptions: (1) steady-state approximation, which assumes that the star is in hydrostatic balance and ignores fluid inertia and the possibility of dynamical tides, (2) equatorial orbit of the companion, and (3) no effects from stellar rotation. The last two assumptions in particular are ostensibly invalid in the case of the TOI-1518 system, which contains a hot Jupiter on a misaligned orbit around a rapidly rotating star (see Section 3.2). The fast rotation of the star and the resulting rotational bulge, combined with the spin–orbit misalignment, mean that the tidal bulge raised by the planet traverses regions of the stellar surface with significantly different surface gravities. This is expected to directly affect the tidal response of the star and the corresponding amplitude of the ellipsoidal distortion signal.

Another possible contributor to an unexpected first harmonic phase-curve modulation is the variable stellar irradiation experienced by the planet. This scenario was explored in detail for the case of KELT-9—a similarly misaligned system with an ultra-hot Jupiter around a rapidly rotating star—where it was found to be the primary source of the unusual phase alignment of the measured first harmonic photometric modulation (Wong et al. 2020c). In short, the rapid stellar rotation induces variations in the effective temperature of the planet-facing hemisphere, which cause the planetary thermal emission to change in response to the time-varying insolation. The three-dimensional orientation of TOI-1518’s rotation axis is not known from the available data, preventing us from being able to directly model the relative phasing of this additional irradiation signal (as was done for the KELT-9 system). Nevertheless, we do expect some level of photometric variability at the first harmonic that is due to the planet’s variable dayside temperature, which may bias the photometric mass estimate.
The previous discussion serves as a cautionary tale about the reliability of photometric mass measurements derived from the ellipsoidal distortion signal. The complexities of the stellar tidal response and the possibility of additional contributions from the planet’s thermal emission mean that many systems are susceptible to significant discrepancies between the measured and expected first harmonic amplitudes. Future RV monitoring of this system will improve the precision of the planet’s mass.

5.3. Conclusion

TESS continues to find numerous transiting exoplanet candidates. As these planets are confirmed, some are bound to become interesting case studies for atmospheric characterization. In this paper, we reported the confirmation of an ultrahot Jupiter on a close-in, highly misaligned orbit around TOI-1518. The stellar, planetary, and orbital parameters derived from fitting the TESS light curve, ground-based transit photometry, and SED are listed in Table 1. The photometry displays a clear secondary eclipse signal, as well as phase-synchronized modulations in flux attributed to the day–night brightness contrast of the planet and the tidal distortion of the host star. In addition, we searched for neutral and ionized Fe in the companion’s atmosphere through high-resolution transmission spectroscopy. We detected Fe at high confidence, and also found evidence for Fe$^+$. TOI-1518b is highly inflated, which makes it amenable to intensive atmospheric characterization. The equilibrium temperature of TOI-1518b is in the regime where the planet might exhibit a thermal inversion (Fortney et al. 2008; Lothringer et al. 2018; Gandhi & Madhusudhan 2019; Malik et al. 2019). This, combined with the brightness of the host star, makes TOI-1518b an attractive target for follow-up emission spectroscopy (Nugroho et al. 2020; Pino et al. 2020; Yan et al. 2020).

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Software: Tapir (Jensen 2013), AstroImageJ (Collins et al. 2017), molecfit (Smette et al. 2015), radvel (Fulton et al. 2018), Lightkurve (Lightkurve Collaboration et al. 2018), Helios-K (Grimm et al. 2021), FastChem (Stock et al. 2018), batman (Kreidberg 2015), emcee (Foreman-Mackey et al. 2013), SME (Valenti & Piskunov 1996), exoplanet (Foreman-Mackey et al. 2021), astropy (Astropy Collaboration et al. 2018).

Appendix

Figure 10 depicts a corner plot containing all the astrophysical parameters fitted for in our joint analysis of the TESS light curve and ground-based full-transit photometry (Section 2.4); for clarity, the limb-darkening coefficients for each data set are not shown. The values of the average relative planetary flux $f_p$, planetary atmospheric brightness modulation amplitude $A_{atm}$, and stellar ellipsoidal distortion amplitude $A_{ellip}$ are given in parts-per-million. The phase offset in the planetary phase curve $\delta$ is provided in degrees. Note the significant correlations between the impact parameter $b$, scaled semimajor axis $a/R_\star$, radius ratio $R_p/R_\star$, and $f_p$—a consequence of the grazing nature of the planetary transit.

Figure 11 shows the full-transit light curves collected as part of ground-based follow-up observations, as described in Section 2.3. Each light curve is labeled with the filter used. The best-fit transit model from the joint TESS and ground-based photometric fit is shown in the bottom panels.
Figure 10. Corner plot of parameters involved in the joint TESS and ground-based light-curve fit.
Figure 11. Ground-based light curves of TOI-1518 with full coverage of the primary transits, collected as part of the TESS Follow-up Program. Top: the photometry at the native time resolution (gray points) and binned (colored points). Each panel is labeled by the respective bandpass. The binning interval for the $B$, $g'$-, and $R$-band observations is 7 minutes; a shorter 4 minute bin size is used for the higher-precision $R_{	ext{bg}}$- and $I$-band transits. Bottom: binned, systematics-corrected light curves, with the best-fit transit model from the joint TESS and ground-based photometric fit (Table 1) plotted in black.

ORCID iDs

Samuel H. C. Cabot @ https://orcid.org/0000-0001-9749-6150
Aaron Bello-Arufe @ https://orcid.org/0000-0003-3555-1223
João M. Mendonça @ https://orcid.org/0000-0002-6907-4476
René Tronsgaard @ https://orcid.org/0000-0003-1001-0707
Ian Wong @ https://orcid.org/0000-0001-9665-8429
George Zhou @ https://orcid.org/0000-0002-4891-3517
Lars A. Buchhave @ https://orcid.org/0000-0003-1605-5666
Debra A. Fischer @ https://orcid.org/0000-0003-2221-0861
Keivan G. Stassun @ https://orcid.org/0000-0002-3481-9052
Victoria Antoci @ https://orcid.org/0000-0002-0865-3650
David Baker @ https://orcid.org/0000-0002-2970-0532
Alexander A. Belinski @ https://orcid.org/0000-0003-3469-0989
Björn Benneke @ https://orcid.org/0000-0001-5578-1498
Luke G. Bouma @ https://orcid.org/0000-0002-0514-5538
Jessie L. Christiansen @ https://orcid.org/0000-0002-8035-4778
Karen A. Collins @ https://orcid.org/0000-0001-6588-9574
Simone Hayeg @ https://orcid.org/0000-0001-8072-0590
Jon M. Jenkins @ https://orcid.org/0000-0002-4715-9460
Eric L. N. Jensen @ https://orcid.org/0000-0002-4625-7333
Bob Massey @ https://orcid.org/0000-0001-8879-7138
Kim K. McLeod @ https://orcid.org/0000-0001-9504-1486
David W. Latham @ https://orcid.org/0000-0001-9911-7388
Edward H. Morgan @ https://orcid.org/0000-0003-1447-6344
George Ricker @ https://orcid.org/0000-0003-2058-6662
Boris S. Safonov @ https://orcid.org/0000-0003-1713-3208
Joshua E. Schlieder @ https://orcid.org/0000-0001-5347-7062
Sara Seager @ https://orcid.org/0000-0002-6892-6948
Avi Shporer @ https://orcid.org/0000-0002-1836-3120
Jeffrey C. Smith @ https://orcid.org/0000-0002-6148-7903
Guillermo Torres @ https://orcid.org/0000-0002-5286-0251
Joseph D. Twicken @ https://orcid.org/0000-0002-6778-7552
Roland Vanderspek @ https://orcid.org/0000-0001-6763-6562
Joshua N. Winn @ https://orcid.org/0000-0002-4265-047X

References

Addison, B. C., Knudstrup, E., Wong, I., et al. 2021, arXiv:2104.12078
Agol, E., Luger, R., & Foreman-Mackey, D. 2020, ApJ, 159, 123
Allart, R., Lovis, C., Pino, L., et al. 2017, A&A, 606, A144
Anderson, D. R., Temple, L. Y., Nielsen, L. D., et al. 2018, arXiv:1809.04897
Astropy Collaboration, Price-Whelan, A. M., Sipőcz, B. M., et al. 2018, AJ, 156, 123
Astropy Collaboration, Robitaille, T. P., Tollerud, E. J., et al. 2013, A&A, 558, A33
Bakos, G., Noyes, R. W., Kovács, G., et al. 2004, PASP, 116, 266
Ballard, S., Fabrycky, D., Fressin, F., et al. 2011, ApJ, 743, 200
Ben-Yami, M., Madhusudhan, N., Cabot, S. H. C., et al. 2020, ApJL, 897, L5
Bieryla, A., Collins, K., Beatty, T. G., et al. 2015, AJ, 150, 12
Birkby, J. L., de Kok, R. J., Brogi, M., et al. 2013, MNRAS, 436, L35
Blackman, R. T., Fischer, D. A., Jurgenson, C. A., et al. 2020, ApJ, 159, 238
Borucki, W. J., Koch, D., Basri, G., et al. 2010, Sci, 327, 977
Bourrier, V., Lecavelier des Etangs, A., Hébrard, G., et al. 2015, A&A, 579, A55
Bower, D. J., Kitzmann, D., Wolf, A. S., et al. 2019, A&A, 631, A103
Breuer, J. M., Fischer, D. A., Blackman, R. T., et al. 2020, ApJ, 160, 67
Breuer, J. M., Fischer, D. A., Valentí, J. A., & Piskunov, N. 2016, ApJS, 225, 32
Brogi, M., Snellen, I. A. G., de Kok, R. J., et al. 2012, Natur, 486, 502
Buchhave, L. A., Bakos, G. Á., Hartman, J. D., et al. 2010, ApJ, 720, 1118
von Essen, C., Mallonn, M., Borre, C., et al. 2020, A&A, 639, A34
Wang, S., Jones, M., Shporer, A., et al. 2019, AJ, 157, 51
Weiss, L. M., Marcy, G. W., Petigura, E. A., et al. 2018, AJ, 155, 48
Winn, J. N., Fabrycky, D., Albrecht, S., & Johnson, J. A. 2010, ApJL, 718, L145
Wong, I., Benneke, B., Shporer, A., et al. 2020a, AJ, 150, 104
Wong, I., Kitzmann, D., Shporer, A., et al. 2021, AJ, 162, 127
Wong, I., Knutson, H. A., Kataria, T., et al. 2016, ApJ, 823, 122
Wong, I., Shporer, A., Daylan, T., et al. 2020b, AJ, 160, 155
Wong, I., Shporer, A., Kitzmann, D., et al. 2020c, AJ, 160, 88
Yan, F., Casassayas-Barris, N., Molaverdikhani, K., et al. 2019, A&A, 632, A69
Yan, F., & Henning, T. 2018, NatAs, 2, 714
Yan, F., Pallé, E., Reiners, A., et al. 2020, A&A, 640, L5
Zhou, G., Bakos, G. Á., Bayliss, D., et al. 2019, AJ, 157, 31
Zhou, G., Rodriguez, J. E., Collins, K. A., et al. 2016, AJ, 152, 136