ares. iii. unveiling the two faces of kelt-7 b with hst wfc3

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Abstract

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Abstract

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1. Introduction

With thousands of planets detected during the previous two decades, the study of atmospheres is at the forefront of exoplanet research and spectroscopic observations now probe these worlds in search of molecular features. Such studies are crucial in the pursuit of understanding the diverse nature of exoplanet chemical compositions, atmospheric processes, internal structures, and the conditions required for planetary formation.

In recent years, there has been a surge in transit spectroscopy observations using both space-borne and ground-based facilities, resulting in significant advancements in our understanding of exoplanet atmospheres. This technique has been used for the detection of multiple molecular absorption features, including water (H2O; Tinetti et al. 2007; Tsiaras et al. 2019), methane (CH4; Swain et al. 2008), and ammonia (MacDonald & Madhusudhan 2017), becoming a cornerstone technique in the pursuit of exoplanet characterization. In particular, the Hubble Space Telescope (HST) has been widely used allowing us to characterize the atmospheres of several hot Jupiters (e.g., Deming et al. 2013; Wakeford et al. 2013; Haynes et al. 2015; Kreidberg et al. 2018; Mikal-Evans et al. 2019) and has begun to observe enough planets for population studies to be undertaken (e.g., Sing et al. 2016; Tsiaras et al. 2018).

For hotter planets, HST transmission spectroscopy has provided evidence for absorption at shorter wavelengths. These are generally attributed to the presence of optical absorbers such as titanium oxide (TiO), vanadium oxide (VO), or iron hydride (FeH). These planets include WASP-121b (Mikal-Evans et al. 2019) and WASP-127 b (Skaf et al. 2020). Additionally, high-resolution ground-based observations have detected the presence of a variety of heavy metals in the atmosphere of KELT-9 b (Hoeijmakers et al. 2019) while lower resolution data were used to claim the presence of aluminum oxide (Al2O3) in Wasp-33 b (von Essen et al. 2019). Thermal emission spectra can also be used to characterize the dayside of exoplanets. In particular, hot Jupiters that emit significant and detectable infrared radiation are the ideal candidates for this approach. As thermal emission is sensitive to the temperature—pressure profile (Madhusudhan et al. 2014), it is possible to probe this structure and demonstrate how it can be driven by the presence of TiO in the atmosphere such as in the cases of Wasp-33 b (Haynes et al. 2015) and WASP-76 b (Edwards et al. 2020a).
The presence of these optical absorbers are an important component in current exoplanet atmospheric models, and are predicted to have a significant impact on the overall physics and chemistry of highly irradiated close-in giant planet atmospheres. Work by Fortney et al. (2008) suggests the dayside atmospheres of such planets can be divided into two separate classes. The first class represents atmospheres that are modeled to have significant opacity due to the presence of TiO and VO gases and thus results in, for example, atmospheric temperature inversions (with hot stratospheres). In the second case, atmospheres that do not possess opacity due to TiO and VO are modeled to redistribute absorbed energy more readily resulting in cooler daysides and warmer nightsides.

WASP-103 b, a similar planet of KELT-7 b, has been studied in both emission and transmission by HST, suggesting a thermal inversion on the dayside but with a featureless transmission spectrum (Kreidberg et al. 2018). On the other hand, HST data for WASP-76 b, another ultra-hot Jupiter, also showed a dayside thermal inversion due to the presence of TiO but a larger water feature and no evidence for optical absorbers in the transmission spectrum (Edwards et al. 2020a).

KELT-7 b is a transiting hot Jupiter, with a mass of $1.28 \pm 0.18 M_J$, a radius of $1.533_{-0.046}^{+0.040} R_J$, an orbital period of $2.7347749 \pm 0.0000039$ days, and an equilibrium temperature of $\sim 2048$ K (Bieryla et al. 2015; see Table 1). KELT-7 b, with relatively low surface gravity, high equilibrium temperature, and a bright host star, is an excellent candidate for thorough atmospheric characterization. It has previously been studied in both transmission and emission using the Spitzer Space Telescope’s InfraRed Array Camera (IRAC; Garhart et al. 2020) and a ground-based eclipse was measured by Martioli et al. (2018). Here we present an analysis of this exoplanet using transmission and emission spectroscopy from the HST’s Wide Field Camera 3 (WFC3). These HST observations allow for two complementary insights into the nature of this planet. We also explore the effects of combining the HST data set with those from Spitzer IRAC and the Transiting Exoplanet Survey Satellite (TESS).

### Table 1: Target Parameters Used in this Study

| Stellar Parameters | Value |
|--------------------|-------|
| Spectral type      | F     |
| Effective temperature (K) | $6789_{-40}^{+50}$ |
| [Fe/H]             | $0.139_{-0.081}^{+0.075}$ |
| Surface gravity (cgs) | $4.149_{-0.019}^{+0.019}$ |
| Radius ($R_\text{J}$) | $1.732_{-0.045}^{+0.045}$ |

| Planetary Parameters | Value |
|----------------------|-------|
| Period (days)        | $2.7347749_{-0.0000039}^{+0.0000039}$ |
| Inclination (deg)    | $83.76_{-0.37}^{+0.38}$ |
| Mass ($M_\text{J}$)  | $1.28_{-0.18}^{+0.18}$ |
| Radius ($R_\text{J}$) | $1.533_{-0.047}^{+0.046}$ |
| Equilibrium temperature (K) | $2048_{-27}^{+27}$ |
| $T_0$ (BJD$_\text{TDB}$) | $2456355.229809_{-0.000198}^{+0.000198}$ |
| $a/R_\text{*}$       | $5.49_{-0.11}^{+0.12}$ |

| Reference           | Bieryla et al. (2015) |

2. Data Analysis

2.1. Observational Data—HST

We obtained the raw spectroscopic observation data from the Mikulski Archive for Space Telescopes (MAST). Both observations of KELT-7 b were undertaken in 2017 as part of Hubble proposal 14767 led by David Sing. These images are the result of a two visits of the target, each containing five HST orbits, using the infrared detector, the WFC3 G141 grism, and a scan rate of $0.9 \text{ s}^{-1}$. Each image consists of five nondestructive reads with an aperture size of $266 \times 266$ pixels in the SPARS10 mode, resulting in a total exposure time of $22.317 \text{ s}$, a maximum signal level of $33,000 \text{ e}^{-}$ per pixel, and a total scan length of about $25^\circ 605$. The spectral data used in this study are listed in the Table 4.

2.2. Reduction and Extraction—Iraclis

The data reduction and calibration was performed using the open-source software Iraclis (Tsiaras et al. 2016b). Our Iraclis analysis starts with the raw spatially scanned spectroscopic images. The data reduction and correction steps are performed in the following order: zero-read subtraction, reference pixel correction, nonlinearity correction, dark current subtraction, gain conversion, sky background subtraction, calibration, flat-field correction, bad pixels, and cosmic ray correction. A detailed description of the data reduction process could be found in Section 2 of Tsiaras et al. (2016b).

Following the reduction process, the flux was extracted from the spatially scanned spectroscopic images to create the final transit light curves per wavelength band. We considered one broadband (white) light curve covering the whole wavelength range in which the G141 grism is sensitive (1.088–1.68 $\mu$m) and spectral light curves with a resolving power of 70 at 1.4 $\mu$m. The bands of the spectral light curves are selected such that the signal-to-noise ratio (S/N) is approximately uniform across the planetary spectrum. We extracted our final light curves from the differential, nondestructive reads. Prior to light curve fitting, we choose to discard the first HST orbit of each visit, as these exhibit much stronger hooks than subsequent orbits (Deming et al. 2013; Zhou et al. 2017).

Our white and spectral light curves were fit using the literature values from Table 2, with only two free parameters: the planet-to-star radius ratio and the mid-transit time. This is motivated by the Earth obscuration gaps, which often means that the ingress and egress of the transit are missed, limiting our ability to refine the semimajor-axis-to-star radius ratio, the inclination, and the eccentricity. The limb-darkening coefficients were selected from the quadratic formula by Claret (2000), and using the stellar parameters in Table 1 and the ATLAS stellar models (Kurucz 1970; Espinoza & Jordán 2015; Morello et al. 2020). Figure 1 shows the raw white light curve, the detrended white light curve, and the fitting residuals for both observations while Figure 2 shows the fits of spectral light curves for each wavelength.

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17 https://archive.stsci.edu/
2.3. Atmospheric Characterization—TauREx3

The reduced spectra obtained using Iraclis were thereafter fitted using TauREx3 (A. F. Al-Refaie et al. 2019, in preparation), a publicly available\(^\text{18}\) Bayesian retrieval framework. For the star parameters and the planet mass, we used the values from Bieryla et al. (2015) listed in Table 1. In our runs we assumed that KELT-7 b possesses a primary atmosphere with a ratio of $\mathrm{H_2}/\mathrm{He} = 0.17$. For the opacity sources, we used the line lists from the ExoMol project (Tennyson et al. 2016), along with those from the HIgh-resolution TRANsmission molecular absorption database (HITRAN; Rothman et al. 1987, 2010). In this publication, we considered six trace gases: $\mathrm{H_2O}$ (Polyansky et al. 2018), CO (Li et al. 2015), TiO (McKemmish et al. 2019), VO (McKemmish et al. 2016), FeH (Dulick et al. 2003; Wende et al. 2010), and $\mathrm{H^+}$ (John 1988; Edwards et al. 2020a). For $\mathrm{H^+}$, the absorption depends only on the mixing ratios of neutral hydrogen atoms and free electrons. As this is a degenerate problem, we fixed the neutral hydrogen volume mixing ratio and imposed a profile inspired from Parmentier et al. (2018) using the two-layer model from Changabat et al. (2019). The only remaining free parameter is $\log(e^-)$. Clouds are modeled assuming a fully opaque gray opacity model.

In this study we use the plane-parallel approximation to model the atmospheres, with pressures ranging from $10^{-2}$ to $10^{6}$ Pa, uniformly sampled in log-space with 100 atmospheric layers. We included the Rayleigh scattering and the collision induced absorption (CIA) of $\mathrm{H_2}$ and $\mathrm{H_2}/\mathrm{He}$ (Abel et al. 2011, 2012; Fletcher et al. 2018). Constant molecular abundance profiles were used, and allowed to vary freely between $10^{-12}$ and $10^{-1}$ in the volume mixing ratio. For the transit spectra, the planetary radius, which here corresponds to the radius at 10 bar, was allowed to vary between $\pm 50\%$ of the literature value. In emission, we set the planet radius to the best-fit value from our transmission retrieval. The cloud top pressure prior ranged from $10^{-2}$ to $10^{6}$ Pa, in log-uniform scale. For the dayside, we do not consider clouds in the model. In transmission, the temperature–pressure profile was assumed to be isothermal while in emission a three-point profile was used.

Finally, we use Multinest (Feroz & Hobson 2008; Feroz et al. 2009, 2019) with 1500 live points and an evidence tolerance of 0.5 in order to explore the likelihood space of atmospheric parameters.

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\(^{18}\) https://github.com/ucl-exoplanets/TauREx3_public
To help contextualize our free retrieval results, we computed a self-consistent forward model with petitCODE, a 1D pressure–temperature iterator solving for radiative-convective and chemical equilibrium (Mollière et al. 2015, 2017). The code includes opacities for H₂, H⁻, H₂O, CO, CO₂, CH₄, HCN, H₂S, NH₃, OH, C₂H₂, PH₃, SiO, FeH, Na, K, Fe, Fe⁺, Mg, Mg⁺, TiO, and VO, as well as radiative scattering and CIA by H₂–H₂ and H₂–He. The atmosphere computed with petitCODE is assumed to be cloud-free, but the possibility of condensing refractory species is included in the equilibrium chemistry. Our petitCODE model for KELT-7 b was computed using the stellar and planetary parameters determined by Bieryla et al. (2015). Here, the surface gravity was computed using the planetary mass and radius. Furthermore, an intrinsic temperature of 600 K was adopted, in accordance with its high equilibrium temperature (Thorngren et al. 2019). Finally, a global planetary averaged redistribution of the irradiation was assumed.

2.5. Ephemeris Refinement

Accurate knowledge of exoplanet transit times is fundamental for atmospheric studies. To ensure KELT-7 b can be observed in the future, we used our HST white light curve mid-time, along with data from TESS (Ricker et al. 2014), to update the ephemeris of the planet. TESS data is publicly available through the MAST archive and we use the pipeline from Edwards et al. (2020b) to download, clean, and fit the two-minute cadence data. KELT-7 b had been studied in Sector 19 and after excluding bad data, we recovered nine transits. These were fitted individually with the planet-to-star radius ratio (Rp/Rs) and transit mid-time (Tmid) as free parameters. We note that the ephemeris of KELT-7 b was also recently refined by Garhart et al. (2020) and we also used the mid-times derived in that study.

3. Results

Our analysis uncovers a rich transmission spectrum that is consistent with a cloud-free atmosphere and suggests the
presence of water and dissociated hydrogen, as shown by the posterior distributions in Figure 3. We calculate the Atmospheric Detectability Index (ADI; Tsiaras et al. 2018) to be 16.8 for the transmission spectrum, indicating strong evidence of atmospheric features. The retrieved temperature of around 1400 K for the terminator region is consistent with the expected value given the equilibrium temperature. However, as we are in a temperature regime ($T_{\text{eq}} \approx 2000$ K) where 3D effects across the limb could occur, we may be biased on the abundances and temperature retrieved (Pluriel et al. 2020).

In contrast, the extracted emission spectrum does not contain strong absorption or emission features. Although the data is not consistent with a simple blackbody, as shown in Figure 4, it can be explained by a varying temperature–pressure profile and H$. We calculate the ADI against a simple blackbody to be 19.6, highlighting the poorness of the

Figure 3. KELT-7 b atmospheric retrieval posterior distributions of the transmission spectrum and the best-fit model with 3$\sigma$ confidence in blue while the HST WFC3 data is represented by the black data points.
blackbody fit and indicating the presence of a detectable atmosphere. However, it could also indicate that 2D effects, such as those suggested by Wilkins et al. (2014), are affecting the observed spectrum. In their study of CoRoT-2 b they observed an emission spectrum similar to that of KELT-7 b (i.e., one that is poorly fit by a blackbody but that can be better fit using two blackbodies). Such inhomogeneities will certainly be important in the analysis of emission data from future missions (Taylor et al. 2020). Our best fit favors a thermal inversion in the stratosphere of KELT-7 b. Although the lower part of the atmosphere has large temperature uncertainties (between 1000 and 3000 K), the middle and top temperature–pressure points are well constrained enough to indicate a thermal inversion with a temperature at the top

Figure 4. KELT-7 b atmospheric retrieval posterior distributions of the emission spectrum and the best fit retrieved with 3σ confidence in red while the HST WFC3 data is represented by the black data points. Also shown is a blackbody fit (green) that converged to a temperature of $T_{\text{BB}} \approx 2500$ K and does not well describe the observations.
between 2900 and 3700 K as shown in Figure 5. Both transmission and emission spectra, along with their best-fit solutions are shown in Figures 3 and 4.

The retrieved emission temperature–pressure profile closely agrees with the petitCODE simulation, both showing a temperature inversion as shown in Figure 5. The retrieved water abundance in transmission is consistent with the predictions although the 1σ upper bound in emission is below what is expected. For TiO, VO, and FeH, the upper bound place on their abundances is significantly above the amount expected from our petitCODE simulations, suggesting the HST WFC3 data is no sensitive enough for us to conclude on their presence/absence.

The transits of KELT-7 b from HST and TESS were seen to arrive early compared to the predictions from Bieryla et al. (2015). As the observation did not include ingress or egress, the HST transit fitting is not as precise, and potentially not as accurate, as the TESS data which captured the whole event. Using this new data, we determined the ephemeris of KELT-7 b to be $P = 2.73476541 \pm 0.00000036$ days and $T_0 = 2458384.425577 \pm 0.000099$ BJD$_{TDB}$ where $P$ is the planet’s period, $T_0$ is the reference mid-time of the transit, and BJD$_{TDB}$ is the barycentric Julian date in the barycentric dynamical. Our derived period is 0.84 s and 0.063 s shorter than the periods from Bieryla et al. (2015) and Garhart et al. (2020) respectively. The ~10 minute residual of the TESS transits from the ephemeris of Bieryla et al. (2015) show the necessity to regularly follow-up transiting planets and for programs such as ExoClock to which our observations have been uploaded. By the launch of the Atmospheric

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**Figure 5.** Results of our self-consistent petitCODE model for KELT-7 b and our retrievals on WFC3 data. Top left: comparison of the temperature–pressure profiles. The petitCODE model (black) features a thermal inversion at 1 mbar, because of absorption by TiO and VO, and closely matches the retrieved profile. Top right: molecular abundances for the petitCODE simulation. The equilibrium fractions of most molecules remain approximately constant for pressures higher than 1 mbar, but drop quickly at lower atmospheric pressures due to thermal dissociation. Bottom left: comparison of constrained molecular abundances in transmission (dotted lines) to those from the petitCODE simulation (solid lines). The water abundance is seen to be within 1σ of that which is predicted. Bottom right: comparison of upper bounds placed on molecular abundances in emission to those from the petitCODE simulation. The 1σ upper bound on the water abundance is significantly lower than the petitCODE simulations. In both transmission and emission, the 1σ upper bounds on TiO, VO, and FeH are well above the predicted abundances, suggesting that the data is not sufficient to comment on their presence/absence.

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19 https://www.exoclock.space
Remote-sensing Infrared Exoplanet Large-survey (ARIEL), in late 2028, around 1400 orbits of KELT-7 b will have occurred since the $T_0$ derived here and the difference in the predicted transit times between the ephemeris from this work and from Garhart et al. (2020) would be around 5 minutes. The observed minus calculated residuals, along with the fitted TESS light curves are shown in Figure 6 while the fitted mid-times can be found in Table 3.

4. Discussion

In transmission, $\text{H}_2\text{O}$ and $\text{H}^+$, ($\log(\text{H}_2\text{O}) = -4.34^{+1.41}_{-1.38}$ and $\log(\text{H}^+) = -4.26^{+1.40}_{-2.62}$) are well defined. As shown in the posterior distributions in Figure 3, correlations exist between the abundances of the molecules, particularly between $\text{H}_2\text{O}$ and $\text{H}^+$. We could expect to also find CO in such a hot atmosphere ($T_{\text{eq}} \approx 2000$ K), evidence for which has been seen for other hot Jupiters: WASP-121b (Evans et al. 2017; Parmentier et al. 2018) and WASP-33b (Haynes et al. 2015). However our retrieval analysis on HST data only provides no evidence for the presence of these molecules. We also explored the addition of Spitzer/IRAC and TESS data, see Section 4.1 for more details.

The nondetection of TiO and VO in the terminator region agreed with predictions from Spiegel et al. (2009). Their work suggests that in highly irradiated atmospheres, similar to KELT-7 b, TiO and VO would be likely to rain out in cold traps and disappear from the visible atmosphere. Observational evidence for these processes was reported in the emission spectrum of Kepler-13Ab (Beatty et al. 2017). To overcome these effects, large advective mixing and higher temperatures (higher than 1800 K) would be required, leading to a large abundance of VO in the dayside of the planet. However, in our analysis, we do not find evidence for VO or TiO on the dayside of KELT-7 b, which could be unraveled with higher S/N in future observations such as ARIEL or the James Webb Space Telescope (JWST).

Also, our analysis potentially finds a large difference in the day/night temperatures with thermal inversion on the dayside, which would indicate that day- and nightside of the planet are poorly coupled by large-scale dynamical processes, thus preventing VO and TiO from reaching the cold nightside and condensating. Fortney et al. (2008) postulated that the presence of optical absorbers would lead, and require, large day–night temperature contrasts, which could suggest that TiO and VO are present but simply beyond the sensitivity of the data.

4.1. Addition of Spitzer Data

KELT-7 b has also been studied, in both transmission and emission, by the Spitzer Space Telescope using each of the 3.6 and 4.5 μm channels of the IRAC. Combining data from multiple instruments or observatories has become the standard within the field as a way of increasing the spectral coverage, seeking to break the degeneracies that occur when fitting data over a narrow wavelength range (e.g., Sing et al. 2016). However, such a procedure is fraught with risk due to potential incompatibilities between data sets. First, it has been shown using different orbital parameters ($a/R_s$ and $i$) in the fitting of the data can lead to offsets between the data sets (Yip et al. 2019). Second, the choice of limb-darkening coefficients can cause vertical shifts in the spectrum.
(Tsiaras et al. 2018). Third, stellar variability, activity, or spots can also induce offsets in the recovered spectra (e.g., Bruno et al. 2020). Finally, imperfect correction of instrument systematics can alter the fitted white light-curve depth, again generating shifts between data sets, making the derived transit/eclipse depths incompatible (e.g., Diamond-Lowe et al. 2014; Stevenson et al. 2014a, 2014b). Each of these effects can of course affect HST data alone but, in this case, the offset would likely only lead to slight changes in the retrieved temperature or radius. When combining data sets, for instance HST WFC3 and Spitzer IRAC, offsets in one of these, or differing offsets in both, could lead to wrongly recovered abundances. In transmission and emission, Spitzer observations are sensitive to CH$_4$ in the 3.6 μm band and CO or CO$_2$ in the 4.5 μm band while WFC3 data cannot constrain these molecules. Thus the detection, or nondetection, would be based entirely off the two Spitzer bands relative to the WFC3 data. An offset in either one of these would instigate the incorrect recovery of the CH$_4$, CO, and CO$_2$ abundances. In emission, these Spitzer points have been used as evidence for the presence of, or lack of, a thermal inversion. Differences in the correction of systematics have led to discrepant results (e.g., HD 209458b; Knutson et al. 2008; Diamond-Lowe et al. 2014).

Here, we tentatively add the Spitzer data for KELT-7 b, taken from the study by Garhart et al. (2020) of 36 hot Jupiters. For the fitting of the Spitzer eclipses, Garhart et al. (2020) froze the orbital parameters to those from Bieryla et al. (2015), overcoming the first hurdle about combining data sets. The transit observations were also fitted with fixed orbital parameters and limb-darkening coefficients from Claret (2000), again showing consistency with our data. The latter two issues, of stellar variability or activity and the detrending of instrument systematics, cannot be easily determined without an overlap in spectral coverage. Therefore, we caution the reader that the compatibility of the data sets cannot be guaranteed. Additionally, we cautiously added the TESS data which was again fitted with same orbital parameters and limb-darkening laws.

The best-fit retrieved spectra, both with and without the Spitzer data, is shown in Figure 7. Little difference is seen between the fits, particularly in transmission where the best fits are almost constantly within 1σ of each other. The recovered temperature–pressure profiles are also practically identical. While this may suggest that the data is compatible, it also shows that the information content of adding Spitzer, in this case, is relatively low. Therefore this begs the question of whether risking data incompatibility is worthwhile when there is little to gain. Figures 8 and 9 show posteriors from the fittings with and without Spitzer, again highlighting the similarity between the fits. In transmission the only noticeable difference is in the recovered CO abundance, which is not constrained in the case of HST alone but the addition of Spitzer suggests an abundance of log(CO) = -4.56$^{+1.72}_{-1.69}$. The second change is in the water abundance

| Wavelength (μm) | Bandwidth (μm) | Transit Depth (%) | Eclipse Depth (%) | Instrument |
|----------------|----------------|-------------------|-------------------|------------|
| 0.8            | 0.4            | 0.7961 ± 0.0018   | 0.03737 ± 0.00145| HST WFC3   |
| 1.12625        | 0.0219         | 0.8159 ± 0.0045   | 0.83951 ± 0.0046 | HST WFC3   |
| 1.14775        | 0.0211         | 0.8134 ± 0.0045   | 0.03669 ± 0.0045 | HST WFC3   |
| 1.16860        | 0.0206         | 0.8062 ± 0.0048   | 0.35444 ± 0.0048 | HST WFC3   |
| 1.18880        | 0.0198         | 0.8152 ± 0.0048   | 0.38764 ± 0.0041 | HST WFC3   |
| 1.20835        | 0.0193         | 0.8195 ± 0.0047   | 0.03787 ± 0.0049 | HST WFC3   |
| 1.22750        | 0.0190         | 0.8181 ± 0.0043   | 0.03865 ± 0.0043 | HST WFC3   |
| 1.24645        | 0.0189         | 0.8114 ± 0.0049   | 0.04363 ± 0.0043 | HST WFC3   |
| 1.26550        | 0.0192         | 0.8172 ± 0.0054   | 0.04558 ± 0.0040 | HST WFC3   |
| 1.28475        | 0.0193         | 0.8170 ± 0.0060   | 0.03292 ± 0.0046 | HST WFC3   |
| 1.30380        | 0.0188         | 0.8130 ± 0.0055   | 0.04558 ± 0.0042 | HST WFC3   |
| 1.32260        | 0.0188         | 0.8237 ± 0.0046   | 0.03523 ± 0.0044 | HST WFC3   |
| 1.34145        | 0.0189         | 0.8163 ± 0.0050   | 0.03577 ± 0.0043 | HST WFC3   |
| 1.36050        | 0.0192         | 0.8174 ± 0.0042   | 0.04638 ± 0.0044 | HST WFC3   |
| 1.38005        | 0.0199         | 0.8117 ± 0.0043   | 0.04633 ± 0.0045 | HST WFC3   |
| 1.40000        | 0.0200         | 0.8171 ± 0.0047   | 0.04750 ± 0.0038 | HST WFC3   |
| 1.42015        | 0.0203         | 0.8191 ± 0.0047   | 0.04555 ± 0.0051 | HST WFC3   |
| 1.44060        | 0.0206         | 0.8127 ± 0.0056   | 0.04515 ± 0.0046 | HST WFC3   |
| 1.46150        | 0.0210         | 0.8179 ± 0.0050   | 0.04580 ± 0.0044 | HST WFC3   |
| 1.48310        | 0.0220         | 0.8164 ± 0.0047   | 0.04480 ± 0.0046 | HST WFC3   |
| 1.50530        | 0.0224         | 0.8006 ± 0.0048   | 0.05031 ± 0.0048 | HST WFC3   |
| 1.52800        | 0.0230         | 0.8112 ± 0.0052   | 0.04825 ± 0.0048 | HST WFC3   |
| 1.55155        | 0.0241         | 0.8052 ± 0.0052   | 0.04765 ± 0.0050 | HST WFC3   |
| 1.57625        | 0.0253         | 0.7989 ± 0.0063   | 0.05397 ± 0.0052 | HST WFC3   |
| 1.60210        | 0.0264         | 0.7872 ± 0.0057   | 0.04457 ± 0.0048 | HST WFC3   |
| 1.62945        | 0.0283         | 0.7936 ± 0.0057   | 0.04332 ± 0.0051 | HST WFC3   |
| 3.6            | 0.75           | 0.7925 ± 0.0062   | 0.1688 ± 0.0046  | Spitzer IRAC |
| 4.5            | 1.015          | 0.8092 ± 0.0036   | 0.1896 ± 0.0057  | Spitzer IRAC |

Note.

$^a$ Taken from Garhart et al. (2020).
recovered in emission, with HST and Spitzer converging to \( \log(H_2O) = -5.11^{\pm 1.37} \) while no constraint can be made in the HST-only case.

On the other hand, the addition of the TESS transmission data drastically changes the solution, removing the detection of dissociated hydrogen, instead preferring FeH to explain the absorption at the shorter wavelengths within the G141 grism due to the shallow TESS transit depth. We also explored the retrieved atmospheric abundances in transmission without the \( H^- \) opacity, again finding that there is little difference when adding Spitzer data as shown in Figure 10. In this case, all data combinations readily agree on the abundances of \( H_2O \), FeH, and CO. However, differences are seen in the evidence for TiO and VO, the presence of which the addition of TESS data of rules out for \( \log(TiO,VO) > -10 \), and in the recovered radius and terminator temperature, with the TESS data set preferring a lower radius and higher temperature. We suggest it is imperative to, at the very least, study multi-instrument data sets separately, as well as combined, when doing model fitting, as we have done in this study.

4.2. Future Characterization

The most effective solution to understanding the source of the absorption seen in transmission between 1.1 and 1.3 \( \mu m \) would be to take more data, namely with the G102 grism of WFC3, which covers 0.8–1.1 \( \mu m \). As there is also archival
Space Telescope Imaging Spectrograph (STIS) transmission data for KELT-7 b, with the G430L and G750L grisms, this would provide continuous coverage from 0.3 to 1.6 μm, allowing for better constraints on the abundances of these optical absorbers. We would also advocate for additional eclipse observations with WFC3, with either grism, to increase the spectral coverage and/or increase the S/N, which may allow for spectral features to be uncovered.

Additionally, future space telescopes JWST (Greene et al. 2016), Twinkle (Edwards et al. 2018), and ARIEL (Tinetti et al. 2018) will provide a far wider wavelength range and these missions will definitively move the exoplanet field from an era
of detection into one of characterization, allowing for the identification of the molecular species present and their chemical profile, insights into the atmospheric temperature profile, and the detection and characterization of clouds. ARIEL, the ESA M4 mission due for launch in 2028, will conduct a survey of $\sim 1000$ planets to answer the question: how chemically diverse are the atmospheres of exoplanets? KELT-7 b has been identified as an excellent target for study with ARIEL (Edwards et al. 2019), through both transmission and emission spectroscopy, and simulated error bars from Mugnai et al. (2020) have been added to the best-fit spectra to showcase this. Figure 11 shows simulated ARIEL and JWST observations and highlights the wavelength coverage performed by those future missions. Additionally ExoWebb (B. Edwards et al. 2020, in preparation) has been used to showcase the capability of JWST for studying this planet.
5. Conclusion

We present spectroscopic transmission and emission observations of KELT-7 b taken with Hubble WFC3. While the transit spectra demonstrates strong absorption features indicative of H$_2$O and H$^-$, the emission spectrum lacks features and can be fitted with CIA alone. We also explore adding data from Spitzer IRAC, with the results being very similar in both transmission and emission. Finally, we find that adding TESS data in our analysis strongly modifies our results. As these instruments do not provide spectral overlap, more data is needed to fully understand the source of the optical absorption seen in transmission. Further observations with Hubble, or with the next generation of observatories, will undoubtedly allow for an enhanced probing of the atmosphere of this intriguing planet. The analysis of archival Hubble data is an essential preparatory step in enriching our comprehension of exoplanet data. 

Figure 10. Posterior distributions for atmospheric retrievals of the transmission spectra of KELT-7 b with various data sets, this time without including H$^-$ as an opacity source. The addition of Spitzer data again has little effect on the retrieved atmospheric properties while TESS data drives the retrieval to higher temperatures and rules out the presence of TiO and VO. All cases require an FeH abundance that is far greater than expected. The reported values for each parameter are those obtained with the fit on all three data sets.
atmospheres, allowing us to begin to appreciate their true diversity and understand the optimal observation strategy for upcoming facilities.

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Software: Iraclis (Tsiaras et al. 2016b), TauREx3 (A. F. Al-Refaie et al. 2019, in preparation), pylightcurve (Tsiaras et al. 2016a), ExoTETHyS (Morello et al. 2020), ArielRad (Mugnai et al. 2020), ExoWebb (B. Edwards et al. 2020, in preparation), Astropy (Astropy Collaboration et al. 2018), h5py (Collette 2013), emcee (Foreman-Mackey et al. 2013), Matplotlib (Hunter 2007), Multinest (Feroz et al. 2009), Pandas (McKinney 2011), Numpy (Oliphant 2011), SciPy (Virtanen et al. 2020).

Appendix

TauREx3 employs Bayesian statistics as the cornerstone for the retrieval analysis (Waldmann et al. 2015a, 2015b). Bayes’ theorem states that

\[
P(\theta|x, M) = \frac{P(x|\theta, M)}{P(x|M)} \times P(\theta|M),
\]

where \(P(\theta|M)\) is the Bayesian prior and \(M\) is the forward model. \(P(\theta|x, M)\) is the posterior probability of the model.
parameters \( \theta \) given the data, \( x \) assuming the forward model \( M \). A Bayesian analysis is implemented in TauREx3 via nested sampling (NS).

TauREx3 includes the implementation of NS Bayesian statistics via Multinest (Feroz & Hobson 2008; Feroz et al. 2009, 2019). NS employs a Monte Carlo approach that constrains via ellipsoids encompassing the parameter space of the highest likelihood. NS determines the Bayesian evidence which is given by

\[
E = \int P(\theta|M)p(x|\theta, M)d\theta
\]

where \( P(x|M) \) is the evidence. These statistical products produced by Multinest are used to perform the best-fit model selection. NS performed by Multinest also allows for efficient parallelization which permits the usage of high performance cluster computing.

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