## Evidence for H2 Dissociation and Recombination Heat Transport in the Atmosphere of KELT-9b

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# Evidence for H2 Dissociation and Recombination Heat Transport in the Atmosphere of KELT-9b

#### **Abstract**

Phase curve observations provide an opportunity to study the energy budgets of exoplanets by quantifying the amount of heat redistributed from their daysides to their nightsides. Theories of phase curves for hot Jupiters have focused on the balance between radiation and dynamics as the primary parameter controlling heat redistribution. However, recent phase curves have shown deviations from the trends that emerge from this theory, which has led to work on additional processes that may affect hot Jupiter energy budgets. One such process, molecular hydrogen dissociation and recombination, can enhance energy redistribution on ultra-hot Jupiters with temperatures above ~2000 K. In order to study the impact of  $H_2$  dissociation on ultra-hot Jupiters, we present a phase curve of KELT-9b observed with the *Spitzer Space Telescope* at 4.5  $\mu$ m. KELT-9b is the hottest known transiting planet, with a 4.5  $\mu$ m dayside brightness temperature of  $4566^{+140}_{-136}$  K and a nightside temperature of  $2556^{+101}_{-97}$  K. We observe a phase curve amplitude of  $0.609 \pm 0.020$  and an offset of  $18.7^{+2.1\circ}_{-2.3}$ . The observed amplitude is too small to be explained by a simple balance between radiation and advection. General circulation models (GCMs) and an energy balance model that include the effects of  $H_2$  dissociation and recombination provide a better match to the data. The GCMs, however, predict a maximum phase offset of  $5^{\circ}$ , which disagrees with our observations at  $>5\sigma$  confidence. This discrepancy may be due to magnetic effects in the planet's highly ionized atmosphere.

Unified Astronomy Thesaurus concepts: Hot Jupiters (753); Exoplanet atmospheres (487)

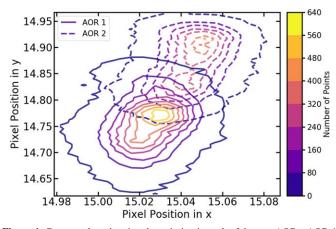
### 1. Introduction

Hot Jupiter phase curve observations have led to a wealth of data on energy transport in highly irradiated planets (Parmentier & Crossfield 2018). This information has spurred the development of theories to describe the resulting trends. The most influential hypothesis has been that the irradiation level is the primary factor controlling energy transport, with hotter planets having shorter radiative timescales and thus less heat redistribution (e.g., Showman & Guillot 2002; Rauscher & Menou 2010; Cowan & Agol 2011; Heng et al. 2011). Lower heat redistribution would lead to increasingly larger phase curve amplitudes and smaller offsets. These trends with irradiation temperature are robust predictions that are born out in models with varying levels of sophistication (e.g., Komacek & Showman 2016).

Recent phase curve observations, however, have shown deviation from these trends, which suggests that the radiative

timescale may not be the only important factor controlling heat redistribution on hot Jupiters (e.g., Zhang et al. 2018; Arcangeli et al. 2019; Keating et al. 2019). In particular, ultra-hot Jupiters with temperatures  $\gtrsim 2000 \, \text{K}$  should have additional important physics because they are hot enough that H<sub>2</sub> dissociates into hydrogen atoms on their daysides and recombines near the terminator (Bell & Cowan 2018; Komacek & Tan 2018; Parmentier et al. 2018). This process is predicted to distribute significant energy in a manner that is similar to latent cooling from water evaporation, with heat deposited in the regions where H recombines into H<sub>2</sub> (Bell & Cowan 2018). Such heat redistribution should lead to smaller phase curve amplitudes (Bell & Cowan 2018; Komacek & Tan 2018). H<sub>2</sub> dissociation also provides a source of hydrogen atoms for the production of H<sup>-</sup>, which is an important opacity source for ultra-hot Jupiters (Arcangeli et al. 2018).

In order to test predictions for energy transport in ultra-hot Jupiters, we present a phase curve of the transiting planet



**Figure 1.** Contour plots showing the pointing in each of the two AORs. AOR 1 is in solid contours and AOR 2 is in dashed contours. The two pointings overlap significantly, allowing for the construction of an accurate pixel sensitivity map spanning the entire observation period.

KELT-9b observed with the *Spitzer Space Telescope* at  $4.5 \mu m$ . KELT-9b is the hottest known planet, with a dayside temperature of  $\sim 4500 \, \mathrm{K}$  (Gaudi et al. 2017). This ultra-hot planet has been shown previously to contain neutral and ionized metals (Hoeijmakers et al. 2018, 2019), and it is predicted to be heavily influenced by H<sub>2</sub> dissociation/recombination (Bell & Cowan 2018; Lothringer et al. 2018; Komacek & Tan 2018; Kitzmann et al. 2018). It is also predicted to be too hot for clouds to form, even on the nightside, which simplifies potential interpretations of its phase curve (Kitzmann et al. 2018). We describe our observations and data reduction process in Section 2. We compare our observations to a set of general circulation models (GCMs) in Section 3 and energy balance models in Section 4. We discuss our results in Section 5.

#### 2. Observations and Data Reduction

We observed a single phase curve of KELT-9b with the InfraRed Array Camera (IRAC) at  $4.5~\mu m$  on 2018 October 22–24 as part of a Cycle 14 large program (program ID: 14059). We used the subarray mode with 0.4~s frame times. Before beginning science observations, we performed a standard 30 minute pre-observation using the PCRS peak-up to mitigate spacecraft drift. Science observations were divided into two contiguous astronomical observation requests (AORs), which lasted for 22.3 and 18.6 hr, respectively. The two AORs had significant overlap in pointing, as shown in Figure 1, and this observation had the most stable pointing overall of the nine phase curves observed to date in program 14059. A total of 371,392 frames were observed. We chose not to analyze the 30 minute pre-observation because it fell on a region of the detector that has little overlap with the two science AORs.

We reduced the data using the Photometry for Orbits, Eclipses, and Transits (POET) pipeline (Campo et al. 2011; Stevenson et al. 2012; Cubillos et al. 2013). We tested a range of fixed and variable aperture sizes (Lanotte et al. 2014) and found that the smallest scatter was achieved with a fixed circular aperture with a radius of 2.5 pixels. We binned sets of four images together for the data reduction because we found that this is the smallest bin size that produces a strong constraint on the point response function full width at half-maximum (PRF FWHM). We modeled position-dependent systematics using Bilinearly Interpolated Subpixel Sensitivity

(BLISS) mapping with a step size of 0.006 pixels (Stevenson et al. 2012). We also decorrelated against the PRF FWHM, as this has been recently shown to improve the fit quality (Mendonça et al. 2018). We tested models with linear, quadratic, and cubic dependences on the PRF FWHM in both the x and y directions, as well as a model without this dependency, and found that a linear model in both directions provides the preferred solution as determined by the Bayesian Information Criterion (BIC). Additionally, we modeled a long-term linear trend over the entire phase curve. We tested a quadratic long-term trend and found that the linear trend is favored, with a  $\Delta$ BIC = 8.

We modeled the phase-dependent emission of KELT-9b using a two-term sinusoid of the form

$$F_p = A_1 \cos\left[\frac{2\pi(t-t_1)}{p}\right] + A_2 \cos\left[\frac{4\pi(t-t_2)}{p}\right],$$
 (1)

where t is time, p = 1.4811 days is the orbital period, and  $A_1$ ,  $A_2$ ,  $t_1$ , and  $t_2$  are free parameters. The second sinusoid allows a fit to an asymmetric phase curve and has been used to model several other phase curves (e.g., Knutson et al. 2012; Stevenson et al. 2017). We tested models using one or three sinusoids, but found a model with two sinusoids is preferred with a  $\Delta BIC = 8$  compared to a one-term model and a  $\Delta BIC = 21$ compared to a three-term model. This test gives us additional confidence in our analysis because third-order harmonics should not exist on a static map (Cowan et al. 2017). We additionally tested for the presence of ellipsoidal variations by fixing the offset  $t_2$  to a time chosen such that the sinusoid has maxima at quadrature and minima at transit and eclipse (Shporer 2017). We found no evidence for ellipsoidal variations above the noise level of the observations, and so left  $t_2$  as a free parameter in the final fit. We fit the transit and eclipses using the model of Mandel & Agol (2002), and used a linear model of stellar limb darkening during the transit.

Wong et al. (2019) found an additional periodicity in *Transiting Exoplanet Survey Satellite (TESS)* phase curves of KELT-9b with a period of  $\approx$ 7.6 hr and semi-amplitude of 117 ppm, which they attribute to stellar pulsations. We confirm the presence of this periodicity through a periodogram analysis of the residuals to our fit. We therefore include a model for this periodicity in our analysis using the equation

$$F_* = 1 + \alpha \sin\left[\frac{2\pi(t - T_0)}{\Pi}\right] + \beta \cos\left[\frac{2\pi(t - T_0)}{\Pi}\right], \quad (2)$$

where  $T_0$  is the transit midpoint, and  $\alpha$ ,  $\beta$ , and  $\Pi$  are free parameters. We find that including these pulsations has an almost negligible influence on our fitted phase curve parameters, which is not surprising because the planet's thermal emission at these infrared wavelengths is more than 10 times larger than the stellar pulsation signal. Nevertheless, we retrieve a period and amplitude for the signal that are consistent with those of Wong et al. (2019), and including it in our model removes some of the correlated noise present in the raw phase curve.

 $<sup>\</sup>overline{^{18}}$  The usefulness of the BIC is limited in this case because BLISS mapping involves several free parameters that are not counted, but the  $\Delta$ BIC can still be used to differentiate between models with different numbers of parameters.

Table 1
Best-fit Values and Errors on All Free Parameters in the Final Fit and Physical
Parameters Derived from the Fit

| Fitted Parameters                          | Value             |  |
|--|-------------------|--|
| Transit Midpoint (BJD <sub>TDB</sub> )     | 2458415.36261(16) |  |
| $R_p/R_s$                                  | 0.08004(41)       |  |
| Linear Limb Darkening, $u_1$               | 0.203(23)         |  |
| Eclipse 1 Midpoint (BJD <sub>TDB</sub> )   | 2458414.62237(32) |  |
| Eclipse 2 Midpoint (BJD <sub>TDB</sub> )   | 2458416.10367(37) |  |
| Eclipse Duration, $T_{14}$ (days)          | 0.16255(47)       |  |
| Eclipse Depth, D (%)                       | 0.3131(62)        |  |
| $A_1$ (ppm)                                | 975(32)           |  |
| $t_1$ (BJD <sub>TDB</sub> )                | 2458414.5544(90)  |  |
| $A_2$ (ppm)                                | 89(22)            |  |
| $t_2$ (BJD <sub>TDB</sub> )                | 2458414.555(33)   |  |
| $\alpha$ (ppm)                             | 0(7)              |  |
| $\beta$ (ppm)                              | -140(20)          |  |
| $\Pi$ (hr)                                 | 7.56(0.13)        |  |
| Linear Ramp, v (ppm/day)                   | 157(36)           |  |
| Linear Fit to $x$ PRF FWHM, PRF $_x$       | -0.1143(43)       |  |
| Linear Fit to y PRF FWHM, PRF <sub>y</sub> | 0.0451(51)        |  |

| Derived Parameters                                       | Value                     |
|--|---------------------------|
| Phase Curve Amplitude, A                                 | $0.609 \pm 0.020$         |
| Phase Offset (°)   | $18.7^{+2.1}_{-2.3}$      |
| Dayside Brightness Temperature, $T_{\text{day}}$ (K)     | $4566^{+140}_{-136}$      |
| Hottest Hemisphere Brightness Temperature (K)            | $4636^{+145}_{-138}$      |
| Nightside Brightness Temperature, $T_{\text{night}}$ (K) | $2556^{+101}_{-97}$       |
| Day–Night Temperature Contrast, $A_T$                    | $0.440^{+0.017}_{-0.016}$ |

**Note.** Numbers in parentheses give  $1\sigma$  uncertainties on the least significant digits. Errors on derived temperatures incorporate the error in the stellar temperatures.

We estimated the parameters using a Differential Evolution Markov Chain Monte Carlo (MCMC) fit (ter Braak & Vrugt 2008) with uniform priors for all parameters. Table 1 lists the values of all fitted parameters. The data exhibit time-correlated noise, so we followed the red noise correction procedure of Diamond-Lowe et al. (2014) and included this effect in our uncertainty estimates using the wavelet analysis described by Carter & Winn (2009). We initially fit for the  $\gamma$  parameter described in Carter & Winn (2009), and then in the final MCMC fixed it to the best-fit value of  $\gamma=0.58$ .

The detrended phase curve is shown in Figure 2. The rms of the residuals when binning the data into 180 points ( $\approx$ 15 minutes/bin) is 118 ppm, and the photon noise is 60 ppm. Table 1 lists several parameters derived from the phase curve, including the dayside and nightside brightness temperatures (4566 $^{+140}_{-136}$  K and 2556 $^{+101}_{-97}$  K, respectively), which were derived using PHOENIX models for the star (Husser et al. 2013). The error on our derived temperatures incorporates the relatively large error on the stellar effective temperature of KELT-9 ( $T_{\rm eff} = 10170 \pm 450$  K; Gaudi et al. 2017). The dayside temperature that we observe at 4.5  $\mu$ m is consistent with the temperature of 4600  $\pm$  150 K observed in the z' band (Gaudi et al. 2017), which is expected from some 1D models of KELT-9b's atmosphere given the measurement uncertainties (Malik et al. 2019). We also derived a day–night temperature

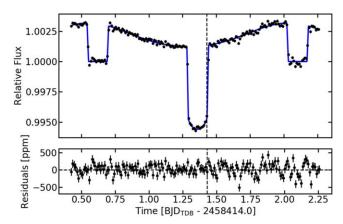


Figure 2. Phase curve of KELT-9b (black points) and best-fit model (blue line). The dashed vertical line shows the break between the two AORs. The lower panel shows residuals to the model fit. The data were binned into sets of four frames for the analysis, but for clarity the data here and in the other figures are shown binned into 180 bins of 2037 frames each ( $\approx$ 15 minutes/bin). The rms of the residuals at this binning is 118 ppm, and the photon noise is 60 ppm.

contrast of

$$A_T = \frac{T_{\text{day}} - T_{\text{night}}}{T_{\text{day}}} = 0.440^{+0.017}_{-0.016},\tag{3}$$

an amplitude of

$$A = \frac{F_{\text{max}} - F_{\text{min}}}{F_{\text{max}}} = 0.609 \pm 0.020 \tag{4}$$

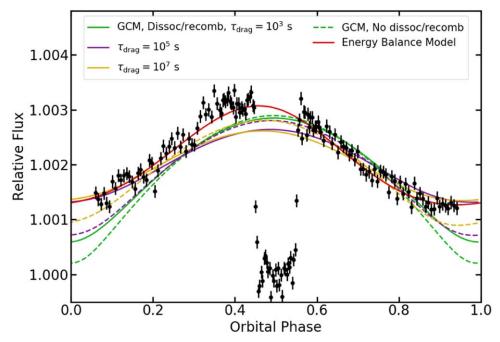
and a phase offset of  $18.7^{+2.1}_{-2.3}$ °.

To ensure the robustness of our results, we tested analyzing the two AORs separately and analyzing a phase curve with the bump in the data at BJD<sub>TDB</sub>  $\approx 2458415.8$  days masked out, and in all cases derived phase offsets and amplitudes that were consistent to within  $1\sigma$ . These data were also analyzed independently by T. Beatty and D. Keating to test for dependence on the data reduction method. The resulting amplitudes and phase offsets agreed within  $1.5\sigma$ . A combined analysis of these data with a *Spitzer*  $3.6~\mu m$  phase curve of KELT-9b will be presented in a future paper (T. Beatty et al. 2019, in preparation).

#### 3. Comparison to GCMs

We used the GCM of Tan & Komacek (2019) to compare the phase curve to numerical predictions. This GCM includes the effects of cooling due to dissociation of molecular hydrogen and heating from recombination of atomic hydrogen, along with changes in the specific heat and specific gas constant due to H<sub>2</sub> dissociation/recombination. The dynamical core of the MITgcm solves the primitive equations of motion on a cubed-sphere grid (Adcroft et al. 2004). We used a double-gray approximation, with one visible and one infrared band in the radiative transfer calculation (Komacek et al. 2017), the opacity of which depends on pressure alone. <sup>19</sup> This opacity profile is the same as that used in Tan & Komacek (2019). We use this simplified opacity profile for our idealized model because relevant opacities have not been calculated exactly at the

<sup>&</sup>lt;sup>19</sup> The thermal opacity profile is  $\log_{10} \kappa_{\text{th}} = 0.0498(\log_{10} p)^2 - 0.1329\log_{10} p - 2.9457$  and the visible opacity profile is  $\log_{10} \kappa_{\text{v}} = 0.0478(\log_{10} p)^2 - 0.1366\log_{10} p - 3.2095$ , with opacity in units of m<sup>2</sup> kg<sup>-1</sup> and pressure is in units of Pa.



**Figure 3.** Phase-folded phase curve of KELT-9b (black data points). The transit at phases of 0 and 1 is omitted to better show the phase variation. Green, purple, and gold lines show GCMs with drag timescales of  $10^3$ ,  $10^5$ , and  $10^7$  s, respectively (Section 3). Solid and dashed lines indicate GCMs with and without the effects of  $H_2$  dissociation and recombination, respectively. The red line shows the EBM including the effects of  $H_2$  dissociation and recombination (Section 4).

temperature of KELT-9b (Freedman et al. 2014) and our GCM setup is unable to fully capture dayside-to-nightside opacity differences.

We used 192 grid points in longitude and 96 in latitude, with 50 vertical levels evenly spaced in log-pressure from 1 mbar to 100 bar. We chose a model top of 1 mbar because the pressure-dependent double-gray opacity scheme used in the GCM does not apply at low pressures (Rauscher & Menou 2012). We fix the stellar  $T_{\rm eff}$ ,  $a/R_*$ , and  $R_p/R_*$  to the values from Gaudi et al. (2017).

We performed multiple GCM experiments with varying frictional drag to crudely represent magnetic effects (Perna et al. 2010; Rauscher & Menou 2013; Rogers & Komacek 2014) and/or large-scale turbulence (Youdin & Mitchell 2010; Fromang et al. 2016). We used a Rayleigh drag that is linear in wind speed,  $f_{\rm drag} = -u/\tau_{\rm drag}$ , where u is velocity and  $\tau_{\rm drag}$  is the frictional drag timescale. We considered a broad range of frictional drag timescales  $\tau_{\text{drag}} = 10^3$ ,  $10^5$ ,  $10^7$  s, to represent the unknown dipolar magnetic field strength (Yadav & Thorngren 2017) and/or length-scale of instabilities (Koll & Komacek 2018). Frictional drag begins to strongly affect the circulation for  $\tau_{\rm drag} \lesssim 10^5\,{\rm s}$  (Komacek et al. 2017), while  $au_{
m drag}=10^7\,{
m s}$  represents very weak drag. For each assumed drag timescale, we ran GCM experiments both including and not including the effects of H<sub>2</sub> dissociation, resulting in six separate GCM experiments. Our simulations with weak drag have an eastward equatorial jet, while our simulations with strong drag have day-to-night flow at photospheric levels. We compare the simulated phase curves to the observations in Figure 3.

We compare our observations to the models using the derived amplitude listed in Table 1. The observed low amplitude  $A=0.609\pm0.020$  indicates significant heat redistribution from the hot dayside. Overall, we find that simulations including the impact of  $H_2$  dissociation/recombination and with relatively weak drag provide a better match to the phase curve amplitude, while those without  $H_2$  dissociation/

recombination and/or with strong drag predict too-large amplitudes and too-cold nightsides.

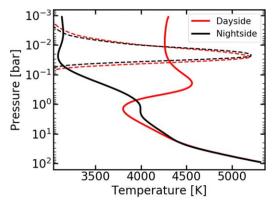
Recent work has suggested that, in many cases, differences in opacity on the daysides and nightsides of hot Jupiters may lead to different pressures being probed through the phase curve in the 4.5  $\mu$ m bandpass (Dobbs-Dixon & Cowan 2017). This can complicate an otherwise straightforward determination of the amount of heat transport in the atmosphere, because the observed day-night temperature contrast may be partially due to the changing photospheric pressure. To determine the impact that this could have on our measurements, we modeled the dayside and nightside emission using the 1D radiative transfer code HELIOS (Malik et al. 2019). We used dayside and nightside temperature-pressure (T-P) profiles from the GCM run with  $\tau_{\text{drag}} = 10^7$  and including the effects of H<sub>2</sub> dissociation/recombination. Figure 4 shows the contribution functions for the *Spitzer* bandpass using these *T–P* profiles. We found that the 4.5  $\mu$ m photosphere was at a pressure of  $\approx$ 10 mbar on both the dayside and the nightside. As the dayside and nightside 4.5  $\mu$ m photospheres are at approximately the same pressure, the temperature difference that we observe is primarily due to horizontal heat transport.

We also observe a large phase offset of  $18.7^{+2.1}_{-2.3}^{\circ}$ . While the GCM experiments including H<sub>2</sub> dissociation and recombination are able to explain the small amplitude of the phase curve, none of the simulations predict the large offset that we observe. The simulations predict an offset of no more than 5°, which is inconsistent with our observations at  $>5\sigma$  confidence.

### 4. Comparison to Energy Balance Models

As a second test of the impact of  $H_2$  dissociation and recombination on the phase curve of KELT-9b, we compare our findings to the open source Bell\_EBM<sup>20</sup> energy balance model (EBM; Bell & Cowan 2018). We use this analytic model

https://github.com/taylorbell57/Bell\_EBM



**Figure 4.** Temperature–pressure profiles from the GCM run with  $\tau_{\rm drag}=10^7$  and including the effects of  $H_2$  dissociation. The dayside and nightside averaged profiles are shown in red and black, respectively. Dashed lines show contribution functions for the center of the *Spitzer 4.5 µm* bandpass from 1D modeling using HELIOS (Malik et al. 2019). The photosphere is at  $\sim$ 10 mbar on both the daysides and nightsides.

in addition to the GCM because it allows us to perform a fit to the data and retrieve parameters that can be compared for models with and without H<sub>2</sub> dissociation/recombination. The EBM was fit to the phase curve using the MCMC package emcee (Foreman-Mackey et al. 2013). In order to allow convergence in a reasonable time frame, we fixed the 4.5  $\mu$ m reference pressure to  $P_0 = 0.1$  bar and fixed the stellar  $T_{\rm eff}$ ,  $a/R_*$ , and  $R_n/R_*$  to the values from Gaudi et al. (2017). We use  $P_0 = 0.1$  bar because it is the approximate depth at which heat is deposited and re-radiated and because longer convergence times at lower pressures mean that it is unfeasible to run even a simplified EBM fit at lower pressures. We fit for the wind speed in the planet's rotating reference frame  $(v_{wind})$  and the planet's Bond albedo  $(A_B)$ . To convert the planet's temperature map into a light curve, we used a 4.5  $\mu$ m stellar brightness temperature of 8287 K found using a PHOENIX stellar model with  $T_{\rm eff} = 10,200 \, \rm K$  (Husser et al. 2013).

Our initial fits showed that the EBM was generally able to recover the phase offset and amplitude of the phase curve, but the fitted phase curve was too sharply peaked, which resulted in an overall poor fit. To improve the fit, we considered another model including a deep redistribution term that redistributes some fraction of the absorbed stellar flux uniformly around the planet. This term mimics the deeper layers (below  $\sim \! 10$  bar) of GCMs, which are nearly longitudinally isothermal as the radiative timescale increases rapidly with depth (e.g., Showman et al. 2009; Rauscher & Menou 2012). This parameter allowed the EBM to fit the data well with a reduced chi-squared of 1.4 for a model with  $\approx \! 21$  % of the absorbed flux redistributed uniformly. The best-fit EBM is shown in Figure 3.

The model including the effects of  $H_2$  dissociation/recombination gives a  $v_{\rm wind}$  of 6.1 km s<sup>-1</sup>, which is on the same order of magnitude as expected for typical ultra-hot Jupiters and is similar to the  $\approx$ 5 km wind speed in our GCM (Koll & Komacek 2018). Meanwhile, neglecting the effects of  $H_2$  dissociation/recombination requires an unphysically high wind speed of 67 km s<sup>-1</sup> to explain the observed heat redistribution, which is further evidence of the impact of  $H_2$  chemistry on the planet's circulation. The model also gives an albedo of 0.195  $\pm$  0.010, which is similar to derived Bond albedos for other ultra-hot Jupiters (Schwartz et al. 2017; Kreidberg et al. 2018; Zhang et al. 2018).

#### 5. Discussion

The most striking result from the KELT-9b phase curve is the small amplitude, which shows the influence of H<sub>2</sub> dissociation/recombination on this planet. Recent work accounting for H<sub>2</sub> dissociation/recombination has demonstrated that the cooling and heating from these processes can transport significant heat, leading to reduced phase curve amplitudes on the hottest ultra-hot Jupiters (Bell & Cowan 2018; Komacek & Tan 2018). When H<sub>2</sub> dissociation is not taken into account, hotter planets are expected to have less heat transport because of their shorter radiative timescales (e.g., Showman & Guillot 2002; Cowan & Agol 2011).

Assuming a solar composition gas and using our model photospheric pressure of  $P \approx 10$  mbar, we estimate that KELT-9b has an extremely short radiative timescale of  $\tau_{\rm rad} \approx 30\,{\rm s}$  (Showman & Guillot 2002). With that short radiative timescale, and ignoring the effects of H<sub>2</sub> dissociation/recombination and frictional drag, the theory of Komacek & Showman (2016) and Zhang & Showman (2017) predicts a normalized dayside-to-nightside temperature contrast of 0.999, which is much greater than the observed value of  $0.440^{+0.017}_{-0.016}$ . Note that including the effects of frictional drag would only act to increase the dayside-to-nightside temperature contrast (Komacek et al. 2017).

This result extends the interpretation of the phase curves of WASP-33b and WASP-103b, two ultra-hot Jupiters that were previously shown to have warm nightsides (Kreidberg et al. 2018; Zhang et al. 2018). These two planets, which both have dayside brightness temperatures around 3000 K, were hypothesized to be impacted by H<sub>2</sub> dissociation/recombination (Bell & Cowan 2018; Komacek & Tan 2018). The extreme irradiation of KELT-9b enhances the impact of H<sub>2</sub> dissociation on the phase curve and provides stronger evidence for this process on ultra-hot Jupiters.

The reduced phase curve amplitude is well fit by both GCMs and the analytic EBM when the effects of  $H_2$  dissociation/recombination are included. We find that relatively weak  $\tau_{\rm drag} \geqslant 10^5$  s is required to match the nightside flux, but strong drag better explains the hot dayside. Additionally, none of the GCMs reproduce the large offset we observe. The large offset could be due to MHD effects that are not currently accounted for in the GCM used in this work. Future work investigating how magnetic effects influence both the phase curve offset and amplitude (e.g., Rogers & Komacek 2014; Rogers 2017; Hindle et al. 2019) could shed light on the remaining discrepancies between the Spitzer observations and GCMs.

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#### References

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Adcroft, A., Hill, C., Campin, J., Marshall, J., & Heimbach, P. 2004, MWRv, 132, 2845

Arcangeli, J., Désert, J.-M., Line, M. R., et al. 2018, ApJL, 855, L30

Arcangeli, J., Désert, J.-M., Parmentier, V., et al. 2019, A&A, 625, A136

Bell, T. J., & Cowan, N. B. 2018, ApJL, 857, L20

Campo, C. J., Harrington, J., Hardy, R. A., et al. 2011, ApJ, 727, 125

Carter, J. A., & Winn, J. N. 2009, ApJ, 704, 51

Cowan, N. B., & Agol, E. 2011, ApJ, 729, 54

Cowan, N. B., Chayes, V., Bouffard, É., Meynig, M., & Haggard, H. M. 2017, MNRAS, 467, 747

Cubillos, P., Harrington, J., Madhusudhan, N., et al. 2013, ApJ, 768, 42

Diamond-Lowe, H., Stevenson, K. B., Bean, J. L., Line, M. R., & Fortney, J. J. 2014, ApJ, 796, 66

Dobbs-Dixon, I., & Cowan, N. B. 2017, ApJL, 851, L26
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Foreman-Mackey, D., Conley, A., Meierjurgen Farr, W., et al. 2013, PASP,
Freedman, R. S., Lustig-Yaeger, J., Fortney, J. J., et al. 2014, ApJS, 214, 25
Fromang, S., Leconte, J., & Heng, K. 2016, A&A, 591, A144
Gaudi, B. S., Stassun, K. G., Collins, K. A., et al. 2017, Natur, 546, 514
Heng, K., Frierson, D. M. W., & Phillipps, P. J. 2011, MNRAS, 418, 2669
Hindle, A., Bushby, P., & Rogers, T. 2019, ApJL, 872, L27
Hoeijmakers, H. J., Ehrenreich, D., Heng, K., et al. 2018, Natur, 560, 453
Hoeijmakers, H. J., Ehrenreich, D., Kitzmann, D., et al. 2019, A&A, 627, A165
Husser, T. O., Wende-von Berg, S., Dreizler, S., et al. 2013, A&A, 553, A6
Keating, D., Cowan, N. B., & Dang, L. 2019, NatAs, in press (doi:10.1038/
  s41550-019-0859-z)
Kitzmann, D., Heng, K., Rimmer, P. B., et al. 2018, ApJ, 863, 183
Knutson, H. A., Lewis, N., Fortney, J. J., et al. 2012, ApJ, 754, 22
Koll, D. D. B., & Komacek, T. D. 2018, ApJ, 853, 133
Komacek, T., Showman, A., & Tan, X. 2017, ApJ, 835, 198
Komacek, T. D., & Showman, A. P. 2016, ApJ, 821, 16
Komacek, T. D., & Tan, X. 2018, RNAAS, 2, 36
Kreidberg, L., Line, M. R., Parmentier, V., et al. 2018, AJ, 156, 17
Lanotte, A. A., Gillon, M., Demory, B. O., et al. 2014, A&A, 572, A73
Lothringer, J., Barman, T., & Koskinen, T. 2018, ApJ, 866, 27
Malik, M., Kitzmann, D., Mendonça, J. M., et al. 2019, AJ, 157, 170
Mandel, K., & Agol, E. 2002, ApJL, 580, L171
Mendonça, J. M., Malik, M., Demory, B.-O., & Heng, K. 2018, AJ, 155, 150
Parmentier, V., & Crossfield, I. J. M. 2018, in Handbook of Exoplanets, ed.
   H. J. Deeg & J. A. Belmonte (Berlin: Springer), 116
Parmentier, V., Line, M. R., Bean, J. L., et al. 2018, A&A, 617, A110
Perna, R., Menou, K., & Rauscher, E. 2010, ApJ, 719, 1421
Rauscher, E., & Menou, K. 2010, ApJ, 714, 1334
Rauscher, E., & Menou, K. 2012, ApJ, 750, 96
Rauscher, E., & Menou, K. 2013, ApJ, 764, 103
Rogers, T., & Komacek, T. 2014, ApJ, 794, 132
Rogers, T. M. 2017, NatAs, 1, 0131
Schwartz, J. C., Kashner, Z., Jovmir, D., & Cowan, N. B. 2017, ApJ, 850, 154
Showman, A. P., Fortney, J. J., Lian, Y., et al. 2009, ApJ, 699, 564
Showman, A. P., & Guillot, T. 2002, A&A, 385, 166
Shporer, A. 2017, PASP, 129, 072001
Stevenson, K. B., Harrington, J., Fortney, J. J., et al. 2012, ApJ, 754, 136
Stevenson, K. B., Line, M. R., Bean, J. L., et al. 2017, AJ, 153, 68
Tan, X., & Komacek, T. D. 2019, ApJ, 886, 26
ter Braak, C. J. F., & Vrugt, J. A. 2008, Statistics and Computing, 18, 435
Wong, I., Shporer, A., Morris, B., et al. 2019, AJ, submitted
Yadav, R., & Thorngren, D. 2017, ApJL, 849, L12
Youdin, A., & Mitchell, J. 2010, ApJ, 721, 1113
Zhang, M., Knutson, H. A., Kataria, T., et al. 2018, AJ, 155, 83
Zhang, X., & Showman, A. 2017, ApJ, 836, 73
```