Water vapour in the atmosphere of the habitable-zone eight-Earth-mass planet K2-18 b

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In the past decade, observations from space and the ground have found water to be the most abundant molecular species, after hydrogen, in the atmospheres of hot, gaseous extrasolar planets\(^1\). Being the main molecular carrier of oxygen, water is a tracer of the origin and the evolution mechanisms of planets. For temperate, terrestrial planets, the presence of water is of great importance as an indicator of habitable conditions. Being small and relatively cold, these planets and their atmospheres are the most challenging to observe, and therefore no atmospheric spectral signatures have so far been detected\(^4\). Super-Earths—planets lighter than ten Earth masses—around later-type stars may provide our first opportunity to study spectroscopically the characteristics of such planets, as they are best suited for transit observations. Here, we report the detection of a spectroscopic signature of water in the atmosphere of K2-18 b—a planet of eight Earth masses in the habitable zone of an M dwarf—with high statistical confidence (Atmospheric Detectability Index\(^5\) = 5.0, \(\sim \)3.6\(\sigma\) (refs. \(^6\)\)). In addition, the derived mean molecular weight suggests an atmosphere still containing some hydrogen. The observations were recorded with the Hubble Space Telescope/Wide Field Camera 3 and analysed with our dedicated, publicly available, algorithms\(^9\). Although the suitability of M dwarfs to host habitable worlds is still under discussion\(^9\), K2-18 b offers an unprecedented opportunity to gain insight into the composition and climate of habitable-zone planets.

Atmospheric characterization of super-Earths is currently within reach of the Wide Field Camera 3 (WFC3) on board the Hubble Space Telescope (HST), combined with the recently implemented spatial scanning observational strategy\(^4\). The spectra of three hot transiting planets with radii less than 3.0 Earth radii (\(R_p\)) have been published so far: Gliese 1214 b\(^15\), HD 97658 b\(^16\) and 55 Cancri e\(^17\). The first two do not show any evident transit depth modulation with wavelength, suggesting an atmosphere covered by thick clouds or made of molecular species heavier than hydrogen, while only the spectrum of 55 Cancri e has revealed a light-weight atmosphere, suggesting hydrogen–helium (H\(_2\)–He) still being present. In addition, the derived mean molecular weight suggests an extended atmosphere or an interior composition with a water (H\(_2\)O) mass fraction lower than 50\% (refs. \(^22\)–\(^24\)).

We analyse here eight transits of K2-18 b, obtained with the WFC3 camera on board the HST. We used our publicly available tools, specialized for HST/WFC3 data\(^6\), to perform the end-to-end analysis from the raw data to the atmospheric parameters. The techniques used here have been validated by the analysis of the largest catalogue of exoplanetary spectra from WFC3\(^3\). Details can be found in Methods, and links to the data and the codes used can be found in ‘Data availability’ and ‘Code availability’, respectively. Along with the data, we provide descriptions of the data structures and instructions on how to reproduce the results presented here. Our analysis resulted in the detection of an atmosphere around K2-18 b with an Atmospheric Detectability Index\(^5\) (ADI; a positively defined logarithmic Bayes factor) of 5.0, or approximately 3.6\(\sigma\) confidence\(^3\), making K2-18 b the first habitable-zone planet in the super-Earth mass regime (1–10 \(M_\oplus\)) with an observed atmosphere around it.

More specifically, nine transits of K2-18 b were observed as part of the HST proposals 13665 and 14682 (principal investigator: Björn Benneke), and the data are available through the Mikulski Archive for Space Telescopes (MAST; see ‘Data availability’). Each transit was observed during five HST orbits, with the G141 infrared grism (1.1–1.7 \(\mu\)m), and each exposure was the result of 16 up-the-ramp samples in the spatial scanning mode. The ninth transit observation suffered from pointing instabilities, and we therefore decided not to include it in this analysis. We extracted the white and the spectral light curves from the reduced images, following our dedicated methodology\(^5\), which has been integrated into an automated, self-consistent and user-friendly Python package named Iraclis (see ‘Code availability’). No systematic variations of the white light curve, \(R_p/R_{\oplus}\), appeared between the eight different observations. This level of stability among the extracted broadband transit depths is not always guaranteed, as consistency problems among different observations emerged in previous analyses\(^9\).

In our analysis, we found that the measured mid-transit times were not consistent with the expected ephemeris\(^1\). We used these results to refine the ephemeris of K2-18 b to be \(P = 32.94007 \pm 0.00003\) days and \(T_0 = 2457363.2109 \pm 0.0004\) BJD\(_{TDB}\) (ref. \(^26\)), where \(P\) is the orbital period of K2-18 b, \(T_0\) is the mid-transit time in BJD\(_{TDB}\), and \(\sigma\) is the uncertainty.
is the period, $T_0$ is the mid-time of the transit and BJD_TDB is the barycentric Julian date in the barycentric dynamical time standard. However, the ephemeris calculated only from the HST data is not consistent with the original detection of K2-18 b. One possibility is that the very sparse data from K2 are not sufficient to give a confident result. Another possibility is that we observe notable transit time variations (TTVs) caused by the other planet in the system, K2-18 c. But more observations over a long period of time are necessary to disentangle the two scenarios. In addition, we used the detrended and time-aligned—that is, with TTVs removed—white light curves to also refine the orbital parameters and found them to be $a/R_*$ = 81.3±1.5 and $i$ = 89.36±0.02 deg, where $a/R_*$ is the orbital semi-major axis normalized to the stellar radius and $i$ is the orbital inclination.

We extracted eight transmission spectra of K2-18 b and combined them, using a weighted average, to produce the final spectrum (Table 1). We interpreted the planetary spectrum using our spectral retrieval algorithm Tau-REx, which combines highly accurate line lists and Bayesian analysis. At an initial stage, we modelled the atmosphere of K2-18 b including all potential absorbers in the observed wavelength range, that is, H2O, carbon monoxide (CO), carbon dioxide (CO2), methane (CH4) and ammonia (NH3). However, we found that only the spectroscopic signature of water vapour is detected with high confidence, so we continued our analysis with only this molecule as trace gas. We modelled the atmosphere following three approaches:

- a cloud-free atmosphere containing only H2O and H2–He
- a cloud-free atmosphere containing H2O, H2–He and molecular nitrogen (N2, which acted as proxy for ´invisible´ molecules not detectable in the WFC3 bandpass but contributing to the mean molecular weight)
- a cloudy (flat-line model) atmosphere containing only H2O and H2–He.

We retrieved a statistically significant atmosphere around K2-18 b in all simulations (Figs. 1–3), and assessed the strength of the

Table 1 | Transit depth ((R_p/R_*)^2) for the different wavelength channels, where $\lambda_1, \lambda_2$ are the lower and upper edge of each wavelength channel, respectively

| $\lambda_1$ (µm) | $\lambda_2$ (µm) | ($R_p/R_*$)^2 (ppm) |
|----------------|----------------|---------------------|
| 1.139 | 1.173 | 2,905±25 |
| 1.173 | 1.204 | 2,939±26 |
| 1.204 | 1.233 | 2,903±24 |
| 1.233 | 1.261 | 2,922±25 |
| 1.261 | 1.289 | 2,891±26 |
| 1.289 | 1.316 | 2,918±26 |
| 1.316 | 1.342 | 2,919±24 |
| 1.342 | 1.369 | 2,955±27 |
| 1.369 | 1.423 | 2,976±25 |
| 1.423 | 1.450 | 2,990±24 |
| 1.450 | 1.478 | 2,895±23 |
| 1.478 | 1.506 | 2,930±23 |
| 1.506 | 1.534 | 2,921±24 |
| 1.534 | 1.562 | 2,875±24 |
| 1.562 | 1.591 | 2,927±25 |
| 1.591 | 1.620 | 2,925±24 |

Fig. 1 | Analysis of the K2-18 b white and spectral light curves, plotted with an offset for clarity. Left: overplotted detrended white (black points) and spectral (coloured points) light curves. Right: overplotted fitted residuals, where $\sigma$ indicates the ratio between the standard deviation of the residuals and the photon noise (see Methods for more details). The black vertical bar indicates the 1,000 ppm scatter level.

Fig. 2 | Best-fit models for the three different scenarios tested. A cloud-free atmosphere containing only H2O and H2–He (blue), a cloud-free atmosphere containing H2O, H2–He and N2 (orange) and a cloudy atmosphere containing only H2O and H2–He (green). Top: best-fit models only. Bottom: 1σ and 2σ uncertainty ranges.
Letters Nature Astronomy

Detection using the ADI, which represents the positively defined logarithmic Bayes factor, where the null hypothesis is a model that contains no active trace gases, Rayleigh scattering or collision-induced absorption—that is, a flat spectrum. The retrieval simulations yield an atmospheric detection with ADI values of 5.0, 4.7, and 4.0, respectively. Such ADI values correspond to a detection of approximately $3.6\sigma$, $3.5\sigma$, and $3.3\sigma$ (refs. 8,9), respectively. This marks the first atmosphere detected around a habitable-zone super-Earth with such a high level of confidence. Although the H$_2$O$^+$/H$_2$–He case appears to be the most favourable, this preference is not statistically significant.

Concerning the composition, retrieval models confirm the presence of water vapour in the atmosphere of K2-18 b in all of the cases studied with high statistical significance. However, it is not possible to constrain either its abundance or the mean molecular weight of the atmosphere. For the H$_2$O$^+$/H$_2$–He case, we found the abundance of H$_2$O to be between 50% and 20%, while for the other two cases, it was between 0.01% and 12.5%. The atmospheric mean molecular weight can be between 5.8 amu and 11.5 amu in the H$_2$O$^+$/H$_2$–He case, and between 2.3 amu and 7.8 amu for the other cases. These results indicate that a non-negligible fraction of the atmosphere is still made of H$_2$–He. Additional trace gases—for example, CH$_4$, NH$_3$,—cannot be excluded, despite not being identified with the current observations: the limiting signal-to-noise ratio and wavelength coverage of HST/WFC3 do not allow the detection of other molecules.

The results presented here confirm the existence of a detectable atmosphere around K2-18 b, making it one of the most interesting known targets for further atmospheric characterization with future observatories such as the James Webb Space Telescope (0.6 μm and 28 μm) and the European Space Agency’s ARIEL (Atmospheric Remote-sensing Infrared Exoplanet Large survey) mission (0.5 μm and 7.8 μm). The wider wavelength coverage of these instruments will provide information on the presence of additional molecular species and on the temperature–pressure profile of the planet, towards studying the planetary climate and potential habitability. Although the subject of habitability for temperate planets around late-type stars is a subject of active discussion, and real progress requires substantially improved observational constraints, the analysis presented here provides the first direct observation of a molecular signature from a habitable-zone exoplanet, connecting these theoretical studies to observations.

Fig. 3 | Posterior distributions for the three different scenarios tested. A cloud-free atmosphere containing only H$_2$O and H$_2$–He (blue), a cloud-free atmosphere containing H$_2$O, H$_2$–He and N$_2$ (orange), and a cloudy atmosphere containing only H$_2$O and H$_2$–He (green). From top to bottom: the volume mixing ratio of H$_2$O, $\log$(H$_2$O); the volume mixing ratio of N$_2$, $\log$(N$_2$); the planetary temperature, $T$, in K; the planetary radius in Jupiter radii ($R_p$); the cloud top pressure, $\log(P_{\text{cloud}})$, in Pa; the derived mean molecular weight, $\mu$, in AMU.
Methods

Observations. Nine transits of K2-18 b were observed as part of the HST proposals 13665 and 14682 (principal investigator: Bjørn Benneke), and the data are available through the MAST. More specifically, the relevant HST visits are: visit 29 (06/12/2015), visit 35 (14/03/2016) and visit 30 (19/05/2016) from proposal 13665; visit 3 (02/12/2016), visit 1 (04/01/2017), visit 2 (06/02/2017), visit 4 (13/04/2017), visit 5 (30/11/2017) and visit 6 (13/05/2018) from proposal 14682. Out of these nine visits, we decided not to include the last one, as it suffered from pointing instabilities.

Each transit was observed during five HST orbits, with the G141 infrared grism of the WFC3 camera (1.1–1.7 µm), in the spatial scanning mode. During an exposure using the spatial scanning mode, the instrument slewed along the cross-dispersion direction, allowing for longer exposure times and increased signal-to-noise ratio, without the risk of saturation. Both forward (increasing row number) and reverse (decreasing row number) scanning were used for these observations. We fitted the light curves using our transit model package PyLightcurve, the nonlinear formula and the stellar parameters in Supplementary Table 2.

We carried out the analysis of the eight K2-18 b transits using our specialized software for the analysis of WFC3 spatially scanned spectroscopic images, which has been integrated into the IRAclis package (see ‘Code availability’). The reduction process included the following steps: zero-read subtraction, reference pixels correction, non-linearity correction, dark current subtraction, gain conversion, sky background subtraction, calibration, flat-field correction and bad-pixels/cosmic-rays correction.

We extracted the white (1.088–1.68 µm) and the spectral (Supplementary Table 1) light curves from the reduced images, taking into account the geometric distortions caused by the tilted detector of the WFC3 infrared channel. The wavelength range of the white light curve corresponds to the edges of the WFC3/G141 throughput (where the throughput drops to 30% of the maximum). In addition, we tested two wavelength grids for the spectral light curves, with resolving powers of 20 and 50. We decided to use the latter as it was able to capture the observed water feature more precisely—that is, there were enough data points within the wavelength range of the water feature to produce a statistically significant result.

We fitted the light curves using our transit model package PyLightcurve, the transit parameters shown in Supplementary Table 2 and limb-darkening coefficients (Supplementary Table 1) calculated based on the PHOENIX model, the nonlinear formula and the stellar parameters in Supplementary Table 2. More specifically, we fitted the white light curves with a transit model (with the planet-to-star radius ratio and the mid-transit time being the only free parameters) along with a model for the systematics. It is common for WFC3 exoplanet observations to be affected by two kinds of time-dependent systematics: the long-term and short-term ramps. The first affects each HST visit and has a linear behaviour, while the second affects each orbit and has an exponential behaviour. The formula we used for the white light curve systematics (\(R_w\)) was the following:

\[
R_w(t) = n_{\text{W}}^\text{sys} \left(1 - r_w(t - T_0)\right) \left(1 - r_w e^{-\frac{\left(t - T_0\right)}{T_1}}\right)
\]

where \(t\) is the time, \(n_{\text{W}}^\text{sys}\) is a normalization factor, \(T_0\) is the mid-transit time, \(r_w\) is the slope of a linear systematic trend along each HST visit and \(T_1\) are the coefficients of an exponential systematic trend along each HST orbit. The normalization factor we used \((n_{\text{W}}^\text{sys})\) was changed to \(n_{\text{W}}^\text{mod}\) for upward scanning directions (forward scanning) and to \(n_{\text{W}}^\text{rev}\) for downward scanning directions (reverse scanning). The reason for using different normalization factors is the slightly different effective exposure time due to the known upstream/downstream effect. We also varied the parameters of the orbit-long exponential ramp for the first orbit in the analysed time series (for \(T_0\) instead of \(r_w\)), because in many other HST observations, the first orbit was affected in a different way from the other orbits. Although we used different ramp parameters from visit to visit, they appear to be consistent, which is an expected behaviour as the number of electrons collected per pixel per second is also consistent.

At the first stage, we fitted the white light curves using the formulae above and the uncertainties per pixel, as propagated through the data reduction process. However, it is common in HST/WFC3 data to have additional scatter that cannot be explained by the ramp model. For this reason, we scaled up the uncertainties per pixel, as propagated through the data reduction process, with the expected epemeris, which we found to be \(P = 3.294007 \pm 0.000003\) days and \(T_0 = 245736.2109 \pm 0.0004\) JD (ref. ). Supplementary Fig. 1 shows the difference between the predicted and the observed transit times using the ephemeris in the literature and the one calculated in this work. We used the detrended and time-aligned—that is, with TTVs removed—white light curves to also refine the orbital parameters (\(\omega, P = 81.3 \pm 1.5\) and \(i = 89.56 \pm 0.02\) deg). At the final stage, we used all of the new parameters (ephemeris, and orbital parameters) to perform a final fit on the white light curves (again with the planet-to-star radius ratio and the mid-transit time being the only free parameters).

Supplementary Fig. 2 shows the raw white light curves, the detrended white light curves, as well as a number of diagnostics, while Supplementary Table 3 presents the fitting results. From these, we can see that

- the final planet-to-star radius ratio is consistent among the eight different transits, demonstrating the stability of both the instrument and the analysis process
- on average, the white light curve residuals show an autocorrelation of 0.17, which is a low number relative to the currently published observations of transiting exoplanets with HST (up to 0.7), indicating a good fit
- uncorrected systematics are still present in the residuals, which, on average, show a scatter two times larger than the expected photon noise
- this extra noise component is taken into account by the increased uncertainties, as the reduced \(\chi^2\) is, on average, 1.16.

Furthermore, we fitted the spectral light curves with a transit model (with the planet-to-star radius ratio being the only free parameter) along with a model for the systematics \((R_w)\) that included the white light curve (divide-white method) and a wavelength-dependent, visit-long slope:

\[
R_w(t) = n_{\text{W}}^\text{sys} \left(1 - r_w(t - T_0)\right) \frac{1 \text{C}_\text{t}}{M_\text{t}}
\]

where \(\chi_0\) is the slope of a wavelength-dependent linear systematic trend along each HST visit, \(\text{C}_\text{t}\) is the white light curve and \(M_\text{t}\) is the best-fit model for the white light curve. Again, the normalization factor we used \((n_{\text{W}}^\text{sys})\) was changed to \(n_{\text{W}}^\text{mod}\) for upward scanning directions (forward scanning) and to \(n_{\text{W}}^\text{rev}\) for downward scanning directions (reverse scanning). Also, in the same way as for the white light curves, we performed an initial fit using the pipeline uncertainties and then refit in order to scale these uncertainties up, for their median to match the standard deviation of the residuals.

Supplementary Figs. 3–19 show the raw spectral light curves, the detrended spectral light curves and the fitting residuals, as well as a number of diagnostics, while Supplementary Table 4 presents all the fitting results and average diagnostics per spectral channel. From these, we can see that

- the spectral light curves residuals show, on average, standard deviations much closer to the photon noise and lower values of autocorrelation, proving the advantage of using the white light curve as a model over the ramp model
- any extra noise component is taken into account by the scaled-up uncertainties, as the reduced \(\chi^2\) is for all channels, on average, close to 1.

Finally, the eight spectra of K2-18 b (Supplementary Fig. 20) were combined, using a weighted average, to produce the final spectrum (Table 1).

Stellar contamination. K2-18 b is a moderately active M2.5V star, with a variability of 1.7% in the B band and 1.38% in the R band. Hence, to make sure that the observed water feature is not the effect of stellar contamination, we fitted the observed spectrum with a model that assumes a flat planetary spectrum and contribution only from the star (M2V star as described in Rackham et al.). The model that best describes our data has a spot coverage of 26% and a faculae coverage of 73%. We plot this spectrum versus the observed one in Supplementary Fig. 21. In addition, we plot the spectrum produced by the spot and faculae combination reported by Rackham et al. and corresponding to a 1% I-band variability, for reference. However, as Supplementary Fig. 21 shows, the best-fit model cannot describe the observed water feature. We therefore conclude that there is no combination of stellar properties that could introduce the observed water feature.

Atmospheric retrieval. We fitted the final planetary spectrum using our Bayesian atmospheric retrieval framework Tau-REx, which fully maps the correlations between the fitted atmospheric parameters through nested sampling.

The atmosphere of K2-18 b was simulated as a plane-parallel atmosphere with pressures ranging from 10^-4 Pa to 10^5 Pa, sampled uniformly in log space by 100 atmospheric layers, assuming an isothermal temperature–pressure profile. We initially tested fitting for a number of trace gases—H₂O (ref. ), CO (ref. ), CO₂ (ref. ), CH₄ (ref. ), NH₃ (ref. )—but found that only water vapour plays a substantial role. Hence, we proceeded with only this molecule. We also included the effect of clouds using a grey/flat-line model, as the quality and wavelength ranges of the currently available observations do not allow us to make any reasonable constraints on the haze properties of the planet. Finally, we included the spectroscopically inactive N₂ as an inactive gas, to account for any unseen absorbers—for example, methane, which is expected at these temperatures. As free parameters in our models, we had the volume mixing ratio of H₂O (log-uniform prior between 10^-10 and 1), the volume mixing ratio of N₂ (log-uniform prior between 10^-8 and 1), the planetary temperature (uniform prior between 260 K and 320 K), the planetary radius (uniform prior between 10^-8 and 10^-7) and the cloud optical depth (uniform prior between 10^-1 and 100).
between 0.05 Jupiter radii ($R_J$) and 0.5 $R_J$ and the cloud top pressure (log-uniform prior between $10^{-4}$ Pa and 100 Pa, where 10 Pa represents a cloud-free atmosphere). We restricted the temperature prior compared with all the possible temperatures for different values of albedo and emissivity because, since we can detect only water, the temperature of the atmospheric part probed must be higher than the freezing point of water (~260 K at 1 mbar).

We identified three solutions: (1) a cloud-free atmosphere containing only H$_2$O and H$_2$. (2) a cloud-free atmosphere containing H$_2$O, H$_2$ and N$_2$. (3) a cloudy atmosphere containing only H$_2$O and H$_2$. The best-fit spectra and the posterior plots are shown in Fig. 3. In all cases, a statistically significant atmosphere around K2-18 b was retrieved with ADIs of 5.0, 4.7 and 4.0, respectively. An ADI of 5 corresponds approximately to a 3.6$\sigma$ detection of an atmosphere. The values are too similar to distinguish between the three scenarios.

### Data availability

The data analysed in this work are available through the NASA MAST HST archive [https://archive.stsci.edu/programmes 13665 and 14682]. The molecular line lists used are available from the ExoMol website (www.exomol.com). The final and intermediate results (reduced data, extracted light curves, light curve fitting results and atmospheric fitting results) are available through the University College London Exoplanets website (https://www.ucl.ac.uk/exoplanets/) and the Open Science Framework (OSF) website at https://doi.org/10.17605/OSF.IO/3TDNJ.

### Code availability

All the software used to produce the presented results are publicly available through the University College London Exoplanets GitHub website (https://github.com/ucl-exoplanets/). More specifically, the codes used were Tau-REx (https://github.com/ucl-exoplanets/TauREx_public), Iraclis (https://github.com/ucl-exoplanets/Iraclis) and PyLightcurve (https://github.com/ucl-exoplanets/pylightcurve).

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### Author contributions

A.T. performed the data analysis and developed the HST analysis software Iracls. LPW developed the atmospheric retrieval software Tau REx; G.T. contributed to the interpretation of the results: J.T. and S.N.Y. coordinated the ExoMol project. All authors discussed the results and commented on the manuscript.

### Competing interests

The authors declare no competing interests.
