Water Vapor Variability in the Thermosphere of Mars during Mars Years 32-35

Zachary Girazian¹

¹The University of Iowa

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Abstract

Using Mars Atmosphere and Volatile Evolution observations, we characterize the variability of water vapor in the Martian thermosphere during Mars Years 32-35. Near a fixed atmospheric pressure level of 5×10^{-7} Pa, the typical water density is $1.3 \ (\pm 0.8) \times 10^3 \ \text{cm}^{-3}$ and the typical water mixing ratio is $10 \ (\pm 6)$ ppm. Thermospheric water levels are higher during the southern spring and summer seasons when Mars is near perihelion and there is significant dust loading in the lower atmosphere. However, the seasonal variation is not the same from year-to-year, likely due annual differences in dust loading. Water vapor is highly correlated with lower atmospheric dust, and increases during both regional and global dust storms. Our results support previous work that found increased dust levels allow more water to be supplied directly to the thermosphere.

 $103~\mathrm{cm}\text{-}3$

Water Vapor Variability in the Thermosphere of Mars during Mars Years 32-35

Z. $Girazian^1$

 $^1\mathrm{Department}$ of Physics and Astronomy, University of Iowa, Iowa City, IA, USA

Key Points:

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- Water vapor in the thermosphere of Mars varies with season.
 - The water abundance is highly correlated with the amount of dust in the lower atmosphere.
 - Water densities increase during regional dust storms.

 $Corresponding \ author: \ Zach \ Girazian, \ \texttt{zachary-girazian@uiowa.edu}$

10 Abstract

Using Mars Atmosphere and Volatile Evolution observations, we characterize the 11 variability of water vapor in the Martian thermosphere during Mars Years 32-35. Near 12 a fixed atmospheric pressure level of $\sim 5 \times 10^{-7}$ Pa, the typical water density is 1.3 (±0.8)× 13 10^3 cm⁻³ and the typical water mixing ratio is 10 (±6) ppm. Thermospheric water lev-14 els are higher during the southern spring and summer seasons when Mars is near per-15 ihelion and there is significant dust loading in the lower atmosphere. However, the sea-16 sonal variation is not the same from year-to-year, likely due annual differences in dust 17 18 loading. Water vapor is highly correlated with lower atmospheric dust, and increases during both regional and global dust storms. Our results support previous work that found 19 increased dust levels allow more water to be supplied directly to the thermosphere. 20

21 Plain Language Summary

Water vapor in the atmosphere of Mars can be transported to high altitudes and 22 ultimately lost to space. Hence, understanding the distribution and variability of water 23 throughout the atmosphere is critical for determining the processes that control water 24 loss at Mars. We present new results on the variability of water vapor in the thermo-25 sphere of Mars (185 km), a region where few water observations exist. We find there 26 is more water during the southern spring and summer seasons when Mars is closest to 27 the Sun and there is a significant amount of dust in the lower atmosphere. We also find 28 that the amount of water is highly correlated with the amount of dust in the lower at-29 mosphere. These results are consistent with previous studies that have shown atmospheric 30 heating from increased dust levels allows more water to be transported to high altitudes 31 without condensing. 32

33 1 Introduction

Water vapor is a common ingredient in the Martian atmosphere, and although it is only a minor constituent, it plays a crucial role in controlling the escape of hydrogen to space. Water is broken apart by ultraviolet sunlight, resulting in H atoms that populate the extended hydrogen exosphere and escape to space via the Jeans mechanism (Hunten & McElroy, 1970; Jakosky, 2021). Hence, understanding the global water cycle is a major theme of research aimed at understanding present-day processes in the atmosphere of Mars, and its evolution through time.

The water vapor is mostly concentrated below ~ 60 km (Aoki et al., 2022) where 41 temperatures are low enough to enable condensation and cloud formation (A. Fedorova 42 et al., 2021), thereby creating a "cold trap" that impedes the upward transport of wa-43 ter. The low-altitude water exhibits a substantial seasonal variation (Montmessin et al., 44 2017; Aoki et al., 2022). Water abundances peak around perihelion and southern sum-45 mer (when the solar longitude, L_s , is between 200° and 300°) when stronger solar heat-46 ing increases the sublimation rate of water frozen in the polar caps (Haberle & Jakosky, 47 1990; Aoki et al., 2022). This period also coincides with dust storm season, which fur-48 ther heats the atmosphere and allows water to propagate to higher altitudes without con-49 densing. Lower atmospheric water abundances reach a minimum during northern spring 50 $(L_s = 0^{\circ}-90^{\circ})$ when solar heating is weakest and atmospheric dust levels are at their min-51 imum. 52

The seasonal trends of low-altitude water are also imprinted on the H escape rate which reaches a maximum during the perihelion and southern summer season (Bhattacharyya et al., 2015; Halekas, 2017; Halekas et al., 2018). Additionally, both global and regional dust storms can rapidly increase the amount of water delivered to the upper atmosphere, leading to impulsive events of increased high altitude water and hydrogen escape (M. Chaffin et al., 2017; A. A. Fedorova et al., 2020; A. Fedorova et al., 2021; Villanueva et al.,
 2021).

Water in the lower and middle atmosphere of Mars has been monitored for several 60 decades (Clancy et al., 1996; A. Fedorova et al., 2006; Smith et al., 2009; A. Fedorova 61 et al., 2021; Crismani et al., 2021). Meanwhile, direct measurements of water vapor at 62 higher altitudes in the thermosphere (100-200 km) have been much more sparse. The 63 most comprehensive study of water vapor in the thermosphere was conducted by Stone 64 et al. (2020) (hereafter, Stone20), who used ion-neutral chemistry to infer water abun-65 dances from measurements by the Mars Atmosphere and Volatile Evolution's (MAVEN's) 66 Neutral Gas and Ion Mass Spectrometer (NGIMS) (Mahaffy et al., 2015; Benna et al., 67 2015). Using the inferred water densities, Stone20 found a seasonal variation in the ther-68 mospheric water abundance that mimics the seasonal variation of water in lower atmo-69 sphere: thermospheric water levels reach a maximum during perihelion and southern sum-70 mer ($L_s = 259^\circ$), and a minimum during northern spring and summer ($L_s = 86^\circ$). Stone20 71 also found a rapid increase in the thermospheric water abundance during two regional 72 dust storms and during the 2018 global dust storm. 73

Given the Stone20 study was completed using MAVEN observations obtained be-74 tween October 2014 and November 2018 (MY32-MY34), we are motivated to extend the 75 analysis with the addition of more recent data. In this study, we use a similar method-76 ology to derive thermospheric water abundances and mixing ratios from NGIMS obser-77 vations. The newly extended data set includes observational coverage throughout Mars 78 Year 35 (MY35), which was unavailable for the Stone 20 study. With the extended data 79 set our study aims to (1) reassess the seasonal cycle of thermospheric water; (2) quan-80 tify the dependence of thermospheric water on lower atmospheric dust content, and (3) 81 determine how the thermospheric water abundance responded during two MY35 regional 82 dust storms. 83

$_{84}$ 2 Method

⁸⁵ NGIMS cannot measure water vapor directly, but measurements of H_3O^+ , and knowl-⁸⁶ edge of its ion-neutral chemistry, allow us to calculate the water vapor abundance. This ⁸⁷ technique was first applied to NGIMS observations by Stone20. We adopt the main as-⁸⁸ pects of their methodology and what follows is only a brief description of the technique ⁸⁹ (we refer readers to Stone20 for the full, detailed description). The technique leverages ⁹⁰ the assumption that injections of water into the thermosphere increase the densities of ⁹¹ protonated ion species such as H_3O^+ , H_2O^+ , HCO^+ (Matta et al., 2013; Fox et al., 2015).

More concretely, in the thermosphere of Mars, H_3O^+ is produced mainly by two chemical reactions, and both involve water:

$$\mathrm{HCO}^{+} + \mathrm{H}_{2}\mathrm{O} \xrightarrow{k_{1}} \mathrm{H}_{3}\mathrm{O}^{+} + \mathrm{CO}$$
 (1)

$$H_2O^+ + H_2O \xrightarrow{k_2} H_3O^+ + OH.$$
 (2)

 H_3O^+ is then lost through dissociative recombination:

$$H_3O^+ + e^- \xrightarrow{\alpha_1} OH + H + H$$
 (3)

$$\xrightarrow{\alpha_2} H_2 O + H \tag{4}$$

$$\xrightarrow{\alpha_3} \text{OH} + \text{H}_2 \tag{5}$$

$$\xrightarrow{\alpha_4} \mathbf{O} + \mathbf{H}_2 + \mathbf{H}.$$
 (6)

Here, k_n are rate coefficients and α_n are dissociative recombination coefficients. Note the production rate of H₃O⁺ is directly proportional to the water vapor abundance (Reactions 1-2). If we assume H₃O⁺ is in photochemical equilibrium, then its production rate is equal to its loss rate. Equating the production and loss rates described in Reactions 1-6 results in the equation

$$[H_2O] \left(k_1 [HCO^+] + k_2 [H_2O^+] \right) = \alpha [H_3O^+] [e^-]$$
(7)

⁹⁹ where brackets indicate abundances, $[e^-]$ is the total electron density, and $\alpha \equiv \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4$. Rearranging Eq. 7 leads to an equation for the water vapor abundance:

$$[H_2O] = \frac{\alpha[H_3O^+][e^-]}{k_1[HCO^+] + k_2[H_2O^+]}$$
(8)

The coefficients are $k_1 = 2.5 \times 10^{-9} \left(\frac{300}{T_i}\right)^{0.5}$ cm³ s⁻¹, $k_2 = 2.1 \times 10^{-9} \left(\frac{300}{T_i}\right)^{0.5}$ cm³ s⁻¹, and $\alpha = 4.36 \times 10^{-7} \left(\frac{300}{T_e}\right)^{0.5}$ cm³ s⁻¹ where T_i is the ion temperature and T_e is the electron temperature, both in units of Kelvin (McElroy et al., 2013). Since H₂O⁺

densities are several orders of magnitude smaller than HCO⁺ densities (Benna et al., 2015),

¹⁰⁵ Equation 8 can be written as

$$[H_2O] = \frac{\alpha[H_3O^+][e^-]}{k_1[HCO^+]}.$$
(9)

¹⁰⁶ Notice the water vapor abundance is proportional to the $[H_3O^+]/[HCO^+]$ density ratio ¹⁰⁷ as discussed in Fox et al. (2015).

All of the abundances on the right hand side of Equation 9 are measured by NGIMS and reported in the Level 2 data products (we assume the electron density is equal to the total ion density). The remaining physical quantities are T_e and T_i , which control the reaction rates. For these, we use Level 2 T_e measurements from MAVEN's Langmuir Probe and Waves (LPW) instrument (Andersson et al., 2015; Ergun et al., 2015) and assume $T_i = T_e$, which is a reasonable assumption at the altitudes of interest (Schunk & Nagy, 2009).

Equation 9 is only valid when H_3O^+ is in photochemical equilibrium. This is sat-115 isfied at lower altitudes in the thermosphere where the pressure is higher and collisions 116 are more frequent. Stone 20 calculated the atmospheric CO_2 density level above which 117 photochemical equilibrium is satisfied. This density level ranges from $\sim 5 \times 10^6$ cm⁻³ 118 - 5×10^7 cm⁻³, depending on the conditions. We adopt a conservative approach to en-119 sure photochemical equilibrium is satisfied by only calculating water abundances at CO_2 120 density levels greater than 5×10^7 cm⁻³ (altitudes below ~ 190 km). We also only con-121 sider dayside observations with solar zenith angles (SZAs) $< 80^{\circ}$. 122

An example of MAVEN observations used to calculate the water abundance dur-123 ing one orbit is shown in Figure 1. The measurements are from the inbound leg of the 124 orbit at 8:30 UTC on 2015-Oct-10. The water abundance profile, derived using the mea-125 sured quantities and Equation 9, is also shown; it only extends up to 183 km (where the 126 CO_2 density drops below 5×10^7 cm⁻³) to ensure H₃O⁺ photochemical equilibrium is 127 satisfied and Equation 9 is valid. The uncertainties of the derived water abundances are 128 assumed to be 70% as estimated by Stone20. Although not shown, the water volume mix-129 ing ratio for each orbit is also calculated by dividing the water abundances by the to-130 tal neutral density measured by NGIMS. 131

As seen in Figure 1, the H_3O^+ densities can be very small ($\leq 10^{-1} \text{ cm}^{-3}$). So small that NGIMS often only registers one or two counts, leading to a high uncertainty in the measured ion density. We mitigate this issue by removing all H_3O^+ densities below a threshold of $1.0 \times 10^{-1} \text{ cm}^{-3}$. This threshold was determined by inspection of the NGIMS Level 1b files, which contain count levels; it roughly corresponds to the threshold above which there are more than ~2 counts.

To derive our final data product for an orbit, we calculate the median water density and mixing ratio at a CO_2 density level between 5×10^7 cm⁻³ to 1×10^8 cm⁻³. This roughly corresponds to a fixed atmospheric pressure level of 5×10^{-7} Pa and an altitude of 185 km. Using this procedure, the water abundance and mixing ratio reported for the orbit shown in Figure 1 are, respectively, $1.8(\pm 1.3) \times 10^3$ cm⁻³ and $20.0(\pm 14.0)$. The median values from each orbit are assigned SZAs, latitudes, longitudes, and local times by averaging the quantity of interest over the same CO_2 density range.

Using this procedure, we derive water abundances and mixing ratios for 1608 or-145 bits using observations from 2015-Jan-1 to 2022-Feb-17. From 2014-2018, MAVEN was 146 in a 4.5 hour elliptical orbit with a periapsis altitude near 160 km. The spacecraft un-147 derwent an aerobraking campaign in early 2019 that changed the orbital period to 3.5 148 hrs and raised the periapsis altitude. For an orbit to be used in our analysis it must con-149 tain both neutral and ion measurements from NGIMS. Only the inbound segment of each 150 orbit is used to ensure the NGIMS background level is small (Stone et al., 2018). We also 151 remove any observations obtained when the spacecraft potential was less than -4 volts, 152 which can cause NGIMS to make erroneous measurements. 153



Figure 1. Density and temperature profiles measured by MAVEN on 2015-Oct-10 at 8:30 UTC. The water vapor profile (blue) is derived from Equation 9 using the densities measured by NGIMS and the electron temperatures measured by LPW. The water vapor profile only extends up to 183 km because Equation 9 is only valid at altitudes where H_3O^+ is in photochemical equilibrium.

We also use dust optical depth maps from Montabone et al. (2015) and (Montabone 154 et al., 2020), which are compiled from dust observations by instruments across several 155 Mars missions. The maps provide a continuous measure of the amount of dust in the lower 156 atmosphere, with near-global coverage. Specifically, the maps provide a continuous record 157 of the optical depth of atmospheric absorption at a wavelength of 9.3 μ m, normalized 158 to an atmospheric pressure level of 610 Pa (also called the "column dust optical depth"). 159 The maps are regularly gridded and have complete global coverage (missing locations 160 are estimated using an interpolation technique). We assign a global average dust opti-161 cal depth to each orbit by averaging the global dust map that is closest in time to the 162 observation. This provides a measure of the lower atmospheric dust content for each wa-163 ter measurement. 164

The top two panels in Figure 2 show the complete set of derived water abundances and mixing ratios, with data points colored by local time of the observation. The large data gaps are primarily from times when MAVEN's periapsis was on the nightside and Equation 9 cannot be used to derive water densities. The bottom panel in Figure 2 shows the time series of global dust optical depth maps with red tracks marking the latitudes of the MAVEN observations.

171 **3 Results**

172 **3.1 Overview**

Our derived mixing ratios are systematically larger (by a factor of \sim 3) than those 173 derived by Stone 20. This discrepancy is primarily caused by the different altitude ranges 174 used to derive the water abundances and mixing ratios. In Stone 20, they calculate the 175 values near periapsis, which can vary from $\sim 150-180$ km throughout the mission. We, 176 instead, calculate them at a fixed CO_2 density level which corresponds to higher altitudes 177 around 185 km. At these higher altitudes the atmospheric density is much lower, result-178 ing in increased mixing ratios. Nonetheless, the variations in our derived mixing ratios 179 match well with Stone20. Further, as a check, we calculated mixing ratios near periap-180 sis and derived values consistent with those found by Stone20. Some minor differences 181 between our derived values are also present because of our slightly different methods. In 182 particular, we use measured electron temperatures while Stone20 assumed the electron 183 temperature was equal to the neutral temperature. 184

As an ensemble, the derived water abundances are consistent with a log-normal distribution (with a high end tail). The average (and standard deviation) of the log₁₀ of the water abundances is 3.1 (± 0.3). This corresponds to an average abundance of 1.3 (± 0.8) × 10³ cm⁻³. For the log₁₀ of the mixing ratio, the average is 1.0 (± 0.3) which corresponds to 10 (± 6) ppm. These average values are representative of the typical water abundance in the Martian thermosphere near a pressure level of 5 × 10⁻⁷ Pa.

3.2 Seasonal Trends

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Although their is significant scatter in the data, there appears to be a sinusoidal variation in the water abundances and mixing ratios (Fig. 2), demonstrating the seasonal dependence of water in the thermosphere. To qualitatively guide the eye, we included sinusoidal curves (orange) in the top two panels of Figure 2. These curves are functions of the form

$$\rho = -A\sin\left(L_s\right) + \rho_o \tag{10}$$

¹⁹² where ρ is the water density or mixing ratio, A is the amplitude, L_s is solar longitude, ¹⁹³ and ρ_o is a constant. For the curves in the figure, the parameters A and ρ_o were not de-¹⁹⁴ termined by any fitting procedure, but instead were arbitrarily chosen to simply act as ¹⁹⁵ a guide.

With the sinusoidal curves plotted as a visual aid, Figure 2 shows there is a con-196 sistent seasonal trend in the thermospheric water levels that is stable across more than 197 three full Martian years. Thermospheric water levels reach a maximum around perihe-198 lion ($L_s = 251^\circ$) and southern summer solstice ($L_s = 270^\circ$). Note, however, the incom-199 plete observational coverage prevents us from determining the precise solar longitude of 200 maximum. Nonetheless, we can conclude that the data are consistent with the maximum 201 being sometime during the period that encompasses perihelion, southern summer, and 202 dust season. The data are also consistent with the thermospheric water levels reaching 203 a minimum around northern summer solstice ($L_s = 90^\circ$). This sinusoidal seasonal vari-204 ation is consistent with the results of Stone 20, who used observations through mid-2018 205 (MY34). The addition of more data in our study, which covers an additional Martian 206 year, shows the seasonal cycle is a stable annual trend that likely occurs every martian 207 year. 208

We can further investigate the seasonal trends with Figure 3, which explicitly shows the water density and mixing ratio as a function of solar longitude, with data points colored according to Mars Year. To remove some of the scatter and focus on large scale trends, we also plot medians from each individual grouping of data points. The median value for each grouping is plotted with a thick circle. The error bar represents the interquartile range. Additionally, for more context, the bottom two panels in Fig. 3 show the dust optical depths and latitudes of the individual observations.

The sinusoidal curves shown in the top two panels of Fig. 3 have the form of Equa-216 tion 10, with the best-fit parameters A and ρ_o determined by fitting the grouped-median 217 values after removing observations during significant dust storms (where the dust op-218 tical depth is greater than 0.4). Note that the use of Equation 10 assumes the solar lon-219 gitudes of the minimum and maximum water levels are at 90° and 270° , respectively. The 220 fits were not used to determine the minimum and maximum locations because the large 221 data gaps prohibit a precise determination. Nonetheless, as discussed previously, the data 222 are consistent with these maximum and minimum values. 223

The best-fit parameters for the water density are $A = 115 \pm 90$ and $\rho_o = 1400 \pm$ 224 60. The best-fit parameters for the mixing ratio are $A = 1.3 \pm 0.7$ and $\rho_o = 11.4 \pm$ 225 0.42. The amplitudes, A, imply the water density and mixing ratio oscillate by $\sim 10\%$ 226 around their mean values over a Martian year. Our derived amplitude for the mixing ra-227 tio is smaller than the amplitude derived by Stone21 who found the oscillation around 228 the mean is $\sim 35\%$. Potential explanations for this discrepancy are the use of additional 229 data in our study, and the slightly different methodologies. The relatively small ampli-230 tudes also imply that variations on shorter timescales are quite significant and often more 231 prominent than the background seasonal trend. 232

It is interesting to compare the water levels at fixed seasons across different Mars 233 years. In general, year-to-year water levels are more consistent during the first half of 234 the year when atmospheric dust levels are lower and less variable. This can be seen in 235 Figure 3 by noting the median values are more tightly grouped around the best-fit curve 236 when L_s is less than 170°. Later in the year, when L_s is greater than 170°, the median 237 values are more spread around the best-fit curve. Near $L_s = 240^\circ$, the two median val-238 ues (MY33 and MY35) are highly separated around the best-fit curve, with the MY33 230 water levels being much higher than those in MY35. This large spread may be caused 240 by the different latitudes of the observations (MY33 are from the southern polar lati-241 tudes and MY35 are from the equatorial latitudes) or the atmospheric dust levels be-242 ing higher in MY33 than in MY35. Near $L_s = 340^\circ$, the median values are somewhat 243 244 separated around the best-fit curve and the water level is higher in MY35 than in MY32 and MY33. All three of these observational periods cover equatorial latitudes so a likely 245 explanation for the differences is the dust levels in the atmosphere, which were highest 246 during MY35. 247

Although there is a background seasonal variation in the thermospheric water abundance, the individual data points can deviate significantly from ta sinusoidal curve when observations from several years are combined. This suggest that the seasonal variation can change from year-to-year and that abrupt variations on shorter time scales are also significant. In the next section, we will explore these shorter term variations by exploring how water levels respond to changing atmospheric dust loading.

3.3 Dust Correlation

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Figure 4 (top) shows the explicit correlation between the water mixing ratio and 255 the global averaged dust optical depth (Section 2). The averages of the log10 of the mix-256 ing ratios are plotted after grouping the data into dust optical depth bins of width 0.03. 257 The Pearson correlation coefficient between the bin-averaged mixing ratios and the bin-258 center dust optical depths is 0.95, indicating a strong correlation. From the lowest to high-259 est dust optical depths, the water mixing ratio increases from ~ 9 to ~ 60 (the highest 260 mixing ratios and dust optical depths are from the 2018 global dust storm (Montabone 261 et al., 2020)). Between these two extremes, the water mixing ratio steadily increases in 262 tandem with the dust optical depth, suggesting that atmospheric dust loading is a ma-263 jor driver of the thermospheric water cycle. This is consistent with the observed seasonal variation of the water mixing ratio, as the dust cycle peaks during southern spring and 265 summer when mixing ratios are highest. Further, because atmospheric dust levels dif-266 fer from year-to-year, one might expect the seasonal variation in the thermospheric wa-267 ter level to also change from year-to-year. 268

A linear fit to the log10 of the mixing ratios yields

$$log_{10}(\rho) = 1.4(\pm 0.08)X + 0.82(\pm 0.02) \tag{11}$$

where ρ is the mixing ratio and X is the dust optical depth. The best-fit slope, 1.4(±0.08), implies the relationship between mixing ratio and the dust optical depth can be approximated by a power law with a dependence that follows $\rho \propto 10^{1.4X}$.

The bottom two panels in Figure 4 show the water mixing ratio and local dust op-272 tical depth during two different time periods. Both of these periods from MY 35 are marked 273 with white boxes in the bottom panel of Figure 2 and coincide with the onset of a re-274 gional dust storm. During the first event in June 2020, MAVEN was crossing into the 275 northern hemisphere at low northern latitudes near 30°. The regional dust storm began 276 just after $L_s = 220^\circ$, when the dust optical depth began to rise. As the dust optical depth 277 increases, the water mixing ratio also increases. This dust storm was typical for this time 278 of year, demonstrating that the onset of typical regional dust storms leads to increased 279 thermospheric water. 280

During the second event (bottom panel) in November and December 2020, a "C-281 type" regional dust storm (Kass et al., 2016) began just after $L_s = 315^{\circ}$. MAVEN was 282 crossing back into the southern hemisphere near equatorial latitudes. At dust storm on-283 set, the water mixing ratio increases from ~ 15 to ~ 20 as the dust optical depth increases 284 from 0.25 to 0.35. At $L_s = 325^{\circ}$, the dust begins to dissipate and the mixing ratios de-285 crease back to pre-storm values. Nonetheless, during the onset of both dust events, the 286 water mixing ratio and dust optical depth are strongly correlated. However, each indi-287 vidual dust storm is somewhat unique and the response of thermospheric water might 288 vary depending on the location with respect to the dust storm.

²⁹⁰ 4 Discussion and Conclusions

MAVEN NGIMS observations allow us to infer water abundances and mixing ratios in the upper atmosphere of Mars to characterize their variability across more than three Martian years. At a fixed atmospheric pressure level of $\sim 5 \times 10^{-7}$ Pa, the typical water density is $1.3 \ (\pm 0.8) \times 10^3 \ \text{cm}^{-3}$ and the typical water mixing ratio is $10 \ (\pm 6)$ ppm. This average mixing ratio is a few times larger than that derived by Stone20, likely due to our somewhat different methodologies. Nonetheless, in agreement with Stone20, we find that water vapor is common in the upper atmosphere.

There is a seasonal variation in the thermospheric water, with higher water levels 298 during the second half of the year – which covers southern summer, perihelion, and dust 200 season $(L_s=180^{\circ}-L_s=360^{\circ})$ – and lower levels during the first half of the year during aphe-300 lion and non-dusty season. This is readily apparent when the water density and mixing 301 ratio is plotted as a time series (Figure 2). However, we find it difficult to quantify how 302 the water vapor varies as a function of solar longitude when the combining observations 303 from three different Mars Years (Figure 3). Instead, we find it is easier to quantify how 304 the water vapor varies as a function of the amount of dust in the lower atmosphere. This 305 leads us to conclude that the atmospheric dust level is a strong driver of the thermospheric 306 water. 307

In particular, we find the water vapor levels are consistent with a monotonically 308 increasing dependence on the global dust optical depth (Figure 4). This supports pre-309 vious work that has shown atmospheric heating from increased dust loading allows wa-310 ter to be transported upwards into the thermosphere without condensing (A. A. Fedorova 311 et al., 2020; M. S. Chaffin et al., 2021; Aoki et al., 2022). Our results suggest that this 312 process is the primary driver of thermospheric water vapor variability. Further support-313 ing this idea is that both regional and global dust storms are clear drivers of an increase 314 in thermospheric water . 315

Many questions remain about the variability of water vapor in the thermosphere such as, (1) how is the water vapor distributed with respect to local time and latitude?; (2) does the thermospheric water vapor depend on solar cycle?; and (3) how does the thermospheric water cycle affect the escape rate of hydrogen from Mars? These questions may be difficult to answer because there are limited observations after 2021 when MAVEN's orbit changed and its periapsis was raised to a higher altitude. Nonetheless, future modelling studies may use our results as constraints to explore these questions.

³²³ 5 Open Research

All of the MAVEN data used in this study are available at https://lasp.colorado .edu/maven/sdc/public/pages/datasets/ngims.html. The dust maps are available at http://www-mars.lmd.jussieu.fr/mars/dust_climatology/. The derived data products will be archived on zenodo.org upon aceptance of the manuscript.

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Figure 2. Derived thermospheric water abundances (top) and mixing ratios (middle). The bottom panel shows dust optical depth maps with the locations of the MAVEN observations shown in red. The dust optical depth data does not extend beyond early 2021. The two white boxes mark the regional dust storms that are analyzed in Section 3.3. The orange curves are sine functions that qualitatively demonstrate the seasonal dependence of water in the thermosphere.



Figure 3. Thermospheric water abundance and mixing ratio as a function of solar longitude (top two panels). The values from each orbit are shown with tiny circles, the grouped-median values are shown with large circles, and the error bars encompass the interquartile range. Colors mark the different Mars Years and the black curves are sinusoidal fits to the median values. The bottom two panels show the dust optical depth and latitude, respectively.



Figure 4. The top panel shows the thermospheric water mixing ratio as a function of globalaveraged column dust optical depth. Small circles show the individual data points, large circles show the average values after separating the data into dust optical depth bins. The black line is a best-fit to the binned averages. The middle panel and bottom panel each show the mixing ratio and dust optical depth during a regional dust storm in MY 35.

Water Vapor Variability in the Thermosphere of Mars during Mars Years 32-35

Z. $Girazian^1$

 $^1\mathrm{Department}$ of Physics and Astronomy, University of Iowa, Iowa City, IA, USA

Key Points:

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- Water vapor in the thermosphere of Mars varies with season.
 - The water abundance is highly correlated with the amount of dust in the lower atmosphere.
 - Water densities increase during regional dust storms.

 $Corresponding \ author: \ Zach \ Girazian, \ \texttt{zachary-girazian@uiowa.edu}$

10 Abstract

Using Mars Atmosphere and Volatile Evolution observations, we characterize the 11 variability of water vapor in the Martian thermosphere during Mars Years 32-35. Near 12 a fixed atmospheric pressure level of $\sim 5 \times 10^{-7}$ Pa, the typical water density is 1.3 (±0.8)× 13 10^3 cm⁻³ and the typical water mixing ratio is 10 (±6) ppm. Thermospheric water lev-14 els are higher during the southern spring and summer seasons when Mars is near per-15 ihelion and there is significant dust loading in the lower atmosphere. However, the sea-16 sonal variation is not the same from year-to-year, likely due annual differences in dust 17 18 loading. Water vapor is highly correlated with lower atmospheric dust, and increases during both regional and global dust storms. Our results support previous work that found 19 increased dust levels allow more water to be supplied directly to the thermosphere. 20

21 Plain Language Summary

Water vapor in the atmosphere of Mars can be transported to high altitudes and 22 ultimately lost to space. Hence, understanding the distribution and variability of water 23 throughout the atmosphere is critical for determining the processes that control water 24 loss at Mars. We present new results on the variability of water vapor in the thermo-25 sphere of Mars (185 km), a region where few water observations exist. We find there 26 is more water during the southern spring and summer seasons when Mars is closest to 27 the Sun and there is a significant amount of dust in the lower atmosphere. We also find 28 that the amount of water is highly correlated with the amount of dust in the lower at-29 mosphere. These results are consistent with previous studies that have shown atmospheric 30 heating from increased dust levels allows more water to be transported to high altitudes 31 without condensing. 32

33 1 Introduction

Water vapor is a common ingredient in the Martian atmosphere, and although it is only a minor constituent, it plays a crucial role in controlling the escape of hydrogen to space. Water is broken apart by ultraviolet sunlight, resulting in H atoms that populate the extended hydrogen exosphere and escape to space via the Jeans mechanism (Hunten & McElroy, 1970; Jakosky, 2021). Hence, understanding the global water cycle is a major theme of research aimed at understanding present-day processes in the atmosphere of Mars, and its evolution through time.

The water vapor is mostly concentrated below ~ 60 km (Aoki et al., 2022) where 41 temperatures are low enough to enable condensation and cloud formation (A. Fedorova 42 et al., 2021), thereby creating a "cold trap" that impedes the upward transport of wa-43 ter. The low-altitude water exhibits a substantial seasonal variation (Montmessin et al., 44 2017; Aoki et al., 2022). Water abundances peak around perihelion and southern sum-45 mer (when the solar longitude, L_s , is between 200° and 300°) when stronger solar heat-46 ing increases the sublimation rate of water frozen in the polar caps (Haberle & Jakosky, 47 1990; Aoki et al., 2022). This period also coincides with dust storm season, which fur-48 ther heats the atmosphere and allows water to propagate to higher altitudes without con-49 densing. Lower atmospheric water abundances reach a minimum during northern spring 50 $(L_s = 0^{\circ}-90^{\circ})$ when solar heating is weakest and atmospheric dust levels are at their min-51 imum. 52

The seasonal trends of low-altitude water are also imprinted on the H escape rate which reaches a maximum during the perihelion and southern summer season (Bhattacharyya et al., 2015; Halekas, 2017; Halekas et al., 2018). Additionally, both global and regional dust storms can rapidly increase the amount of water delivered to the upper atmosphere, leading to impulsive events of increased high altitude water and hydrogen escape (M. Chaffin et al., 2017; A. A. Fedorova et al., 2020; A. Fedorova et al., 2021; Villanueva et al.,
 2021).

Water in the lower and middle atmosphere of Mars has been monitored for several 60 decades (Clancy et al., 1996; A. Fedorova et al., 2006; Smith et al., 2009; A. Fedorova 61 et al., 2021; Crismani et al., 2021). Meanwhile, direct measurements of water vapor at 62 higher altitudes in the thermosphere (100-200 km) have been much more sparse. The 63 most comprehensive study of water vapor in the thermosphere was conducted by Stone 64 et al. (2020) (hereafter, Stone20), who used ion-neutral chemistry to infer water abun-65 dances from measurements by the Mars Atmosphere and Volatile Evolution's (MAVEN's) 66 Neutral Gas and Ion Mass Spectrometer (NGIMS) (Mahaffy et al., 2015; Benna et al., 67 2015). Using the inferred water densities, Stone20 found a seasonal variation in the ther-68 mospheric water abundance that mimics the seasonal variation of water in lower atmo-69 sphere: thermospheric water levels reach a maximum during perihelion and southern sum-70 mer ($L_s = 259^\circ$), and a minimum during northern spring and summer ($L_s = 86^\circ$). Stone20 71 also found a rapid increase in the thermospheric water abundance during two regional 72 dust storms and during the 2018 global dust storm. 73

Given the Stone20 study was completed using MAVEN observations obtained be-74 tween October 2014 and November 2018 (MY32-MY34), we are motivated to extend the 75 analysis with the addition of more recent data. In this study, we use a similar method-76 ology to derive thermospheric water abundances and mixing ratios from NGIMS obser-77 vations. The newly extended data set includes observational coverage throughout Mars 78 Year 35 (MY35), which was unavailable for the Stone 20 study. With the extended data 79 set our study aims to (1) reassess the seasonal cycle of thermospheric water; (2) quan-80 tify the dependence of thermospheric water on lower atmospheric dust content, and (3) 81 determine how the thermospheric water abundance responded during two MY35 regional 82 dust storms. 83

$_{84}$ 2 Method

⁸⁵ NGIMS cannot measure water vapor directly, but measurements of H_3O^+ , and knowl-⁸⁶ edge of its ion-neutral chemistry, allow us to calculate the water vapor abundance. This ⁸⁷ technique was first applied to NGIMS observations by Stone20. We adopt the main as-⁸⁸ pects of their methodology and what follows is only a brief description of the technique ⁸⁹ (we refer readers to Stone20 for the full, detailed description). The technique leverages ⁹⁰ the assumption that injections of water into the thermosphere increase the densities of ⁹¹ protonated ion species such as H_3O^+ , H_2O^+ , HCO^+ (Matta et al., 2013; Fox et al., 2015).

More concretely, in the thermosphere of Mars, H_3O^+ is produced mainly by two chemical reactions, and both involve water:

$$\mathrm{HCO}^{+} + \mathrm{H}_{2}\mathrm{O} \xrightarrow{k_{1}} \mathrm{H}_{3}\mathrm{O}^{+} + \mathrm{CO}$$
 (1)

$$H_2O^+ + H_2O \xrightarrow{k_2} H_3O^+ + OH.$$
 (2)

 H_3O^+ is then lost through dissociative recombination:

$$H_3O^+ + e^- \xrightarrow{\alpha_1} OH + H + H$$
 (3)

$$\xrightarrow{\alpha_2} H_2 O + H \tag{4}$$

$$\xrightarrow{\alpha_3} \text{OH} + \text{H}_2 \tag{5}$$

$$\xrightarrow{\alpha_4} \mathbf{O} + \mathbf{H}_2 + \mathbf{H}.$$
 (6)

Here, k_n are rate coefficients and α_n are dissociative recombination coefficients. Note the production rate of H₃O⁺ is directly proportional to the water vapor abundance (Reactions 1-2). If we assume H₃O⁺ is in photochemical equilibrium, then its production rate is equal to its loss rate. Equating the production and loss rates described in Reactions 1-6 results in the equation

$$[H_2O] \left(k_1 [HCO^+] + k_2 [H_2O^+] \right) = \alpha [H_3O^+] [e^-]$$
(7)

⁹⁹ where brackets indicate abundances, $[e^-]$ is the total electron density, and $\alpha \equiv \alpha_1 + \alpha_2 + \alpha_3 + \alpha_4$. Rearranging Eq. 7 leads to an equation for the water vapor abundance:

$$[H_2O] = \frac{\alpha[H_3O^+][e^-]}{k_1[HCO^+] + k_2[H_2O^+]}$$
(8)

The coefficients are $k_1 = 2.5 \times 10^{-9} \left(\frac{300}{T_i}\right)^{0.5}$ cm³ s⁻¹, $k_2 = 2.1 \times 10^{-9} \left(\frac{300}{T_i}\right)^{0.5}$ cm³ s⁻¹, and $\alpha = 4.36 \times 10^{-7} \left(\frac{300}{T_e}\right)^{0.5}$ cm³ s⁻¹ where T_i is the ion temperature and T_e is the electron temperature, both in units of Kelvin (McElroy et al., 2013). Since H₂O⁺

densities are several orders of magnitude smaller than HCO⁺ densities (Benna et al., 2015),

¹⁰⁵ Equation 8 can be written as

$$[H_2O] = \frac{\alpha[H_3O^+][e^-]}{k_1[HCO^+]}.$$
(9)

¹⁰⁶ Notice the water vapor abundance is proportional to the $[H_3O^+]/[HCO^+]$ density ratio ¹⁰⁷ as discussed in Fox et al. (2015).

All of the abundances on the right hand side of Equation 9 are measured by NGIMS and reported in the Level 2 data products (we assume the electron density is equal to the total ion density). The remaining physical quantities are T_e and T_i , which control the reaction rates. For these, we use Level 2 T_e measurements from MAVEN's Langmuir Probe and Waves (LPW) instrument (Andersson et al., 2015; Ergun et al., 2015) and assume $T_i = T_e$, which is a reasonable assumption at the altitudes of interest (Schunk & Nagy, 2009).

Equation 9 is only valid when H_3O^+ is in photochemical equilibrium. This is sat-115 isfied at lower altitudes in the thermosphere where the pressure is higher and collisions 116 are more frequent. Stone 20 calculated the atmospheric CO_2 density level above which 117 photochemical equilibrium is satisfied. This density level ranges from $\sim 5 \times 10^6$ cm⁻³ 118 - 5×10^7 cm⁻³, depending on the conditions. We adopt a conservative approach to en-119 sure photochemical equilibrium is satisfied by only calculating water abundances at CO_2 120 density levels greater than 5×10^7 cm⁻³ (altitudes below ~ 190 km). We also only con-121 sider dayside observations with solar zenith angles (SZAs) $< 80^{\circ}$. 122

An example of MAVEN observations used to calculate the water abundance dur-123 ing one orbit is shown in Figure 1. The measurements are from the inbound leg of the 124 orbit at 8:30 UTC on 2015-Oct-10. The water abundance profile, derived using the mea-125 sured quantities and Equation 9, is also shown; it only extends up to 183 km (where the 126 CO_2 density drops below 5×10^7 cm⁻³) to ensure H₃O⁺ photochemical equilibrium is 127 satisfied and Equation 9 is valid. The uncertainties of the derived water abundances are 128 assumed to be 70% as estimated by Stone20. Although not shown, the water volume mix-129 ing ratio for each orbit is also calculated by dividing the water abundances by the to-130 tal neutral density measured by NGIMS. 131

As seen in Figure 1, the H_3O^+ densities can be very small ($\leq 10^{-1} \text{ cm}^{-3}$). So small that NGIMS often only registers one or two counts, leading to a high uncertainty in the measured ion density. We mitigate this issue by removing all H_3O^+ densities below a threshold of $1.0 \times 10^{-1} \text{ cm}^{-3}$. This threshold was determined by inspection of the NGIMS Level 1b files, which contain count levels; it roughly corresponds to the threshold above which there are more than ~2 counts.

To derive our final data product for an orbit, we calculate the median water density and mixing ratio at a CO_2 density level between 5×10^7 cm⁻³ to 1×10^8 cm⁻³. This roughly corresponds to a fixed atmospheric pressure level of 5×10^{-7} Pa and an altitude of 185 km. Using this procedure, the water abundance and mixing ratio reported for the orbit shown in Figure 1 are, respectively, $1.8(\pm 1.3) \times 10^3$ cm⁻³ and $20.0(\pm 14.0)$. The median values from each orbit are assigned SZAs, latitudes, longitudes, and local times by averaging the quantity of interest over the same CO_2 density range.

Using this procedure, we derive water abundances and mixing ratios for 1608 or-145 bits using observations from 2015-Jan-1 to 2022-Feb-17. From 2014-2018, MAVEN was 146 in a 4.5 hour elliptical orbit with a periapsis altitude near 160 km. The spacecraft un-147 derwent an aerobraking campaign in early 2019 that changed the orbital period to 3.5 148 hrs and raised the periapsis altitude. For an orbit to be used in our analysis it must con-149 tain both neutral and ion measurements from NGIMS. Only the inbound segment of each 150 orbit is used to ensure the NGIMS background level is small (Stone et al., 2018). We also 151 remove any observations obtained when the spacecraft potential was less than -4 volts, 152 which can cause NGIMS to make erroneous measurements. 153



Figure 1. Density and temperature profiles measured by MAVEN on 2015-Oct-10 at 8:30 UTC. The water vapor profile (blue) is derived from Equation 9 using the densities measured by NGIMS and the electron temperatures measured by LPW. The water vapor profile only extends up to 183 km because Equation 9 is only valid at altitudes where H_3O^+ is in photochemical equilibrium.

We also use dust optical depth maps from Montabone et al. (2015) and (Montabone 154 et al., 2020), which are compiled from dust observations by instruments across several 155 Mars missions. The maps provide a continuous measure of the amount of dust in the lower 156 atmosphere, with near-global coverage. Specifically, the maps provide a continuous record 157 of the optical depth of atmospheric absorption at a wavelength of 9.3 μ m, normalized 158 to an atmospheric pressure level of 610 Pa (also called the "column dust optical depth"). 159 The maps are regularly gridded and have complete global coverage (missing locations 160 are estimated using an interpolation technique). We assign a global average dust opti-161 cal depth to each orbit by averaging the global dust map that is closest in time to the 162 observation. This provides a measure of the lower atmospheric dust content for each wa-163 ter measurement. 164

The top two panels in Figure 2 show the complete set of derived water abundances and mixing ratios, with data points colored by local time of the observation. The large data gaps are primarily from times when MAVEN's periapsis was on the nightside and Equation 9 cannot be used to derive water densities. The bottom panel in Figure 2 shows the time series of global dust optical depth maps with red tracks marking the latitudes of the MAVEN observations.

171 **3 Results**

172 **3.1 Overview**

Our derived mixing ratios are systematically larger (by a factor of \sim 3) than those 173 derived by Stone 20. This discrepancy is primarily caused by the different altitude ranges 174 used to derive the water abundances and mixing ratios. In Stone 20, they calculate the 175 values near periapsis, which can vary from $\sim 150-180$ km throughout the mission. We, 176 instead, calculate them at a fixed CO_2 density level which corresponds to higher altitudes 177 around 185 km. At these higher altitudes the atmospheric density is much lower, result-178 ing in increased mixing ratios. Nonetheless, the variations in our derived mixing ratios 179 match well with Stone20. Further, as a check, we calculated mixing ratios near periap-180 sis and derived values consistent with those found by Stone20. Some minor differences 181 between our derived values are also present because of our slightly different methods. In 182 particular, we use measured electron temperatures while Stone20 assumed the electron 183 temperature was equal to the neutral temperature. 184

As an ensemble, the derived water abundances are consistent with a log-normal distribution (with a high end tail). The average (and standard deviation) of the log₁₀ of the water abundances is 3.1 (± 0.3). This corresponds to an average abundance of 1.3 (± 0.8) × 10³ cm⁻³. For the log₁₀ of the mixing ratio, the average is 1.0 (± 0.3) which corresponds to 10 (± 6) ppm. These average values are representative of the typical water abundance in the Martian thermosphere near a pressure level of 5 × 10⁻⁷ Pa.

3.2 Seasonal Trends

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Although their is significant scatter in the data, there appears to be a sinusoidal variation in the water abundances and mixing ratios (Fig. 2), demonstrating the seasonal dependence of water in the thermosphere. To qualitatively guide the eye, we included sinusoidal curves (orange) in the top two panels of Figure 2. These curves are functions of the form

$$\rho = -A\sin\left(L_s\right) + \rho_o \tag{10}$$

¹⁹² where ρ is the water density or mixing ratio, A is the amplitude, L_s is solar longitude, ¹⁹³ and ρ_o is a constant. For the curves in the figure, the parameters A and ρ_o were not de-¹⁹⁴ termined by any fitting procedure, but instead were arbitrarily chosen to simply act as ¹⁹⁵ a guide.

With the sinusoidal curves plotted as a visual aid, Figure 2 shows there is a con-196 sistent seasonal trend in the thermospheric water levels that is stable across more than 197 three full Martian years. Thermospheric water levels reach a maximum around perihe-198 lion ($L_s = 251^\circ$) and southern summer solstice ($L_s = 270^\circ$). Note, however, the incom-199 plete observational coverage prevents us from determining the precise solar longitude of 200 maximum. Nonetheless, we can conclude that the data are consistent with the maximum 201 being sometime during the period that encompasses perihelion, southern summer, and 202 dust season. The data are also consistent with the thermospheric water levels reaching 203 a minimum around northern summer solstice ($L_s = 90^\circ$). This sinusoidal seasonal vari-204 ation is consistent with the results of Stone 20, who used observations through mid-2018 205 (MY34). The addition of more data in our study, which covers an additional Martian 206 year, shows the seasonal cycle is a stable annual trend that likely occurs every martian 207 year. 208

We can further investigate the seasonal trends with Figure 3, which explicitly shows the water density and mixing ratio as a function of solar longitude, with data points colored according to Mars Year. To remove some of the scatter and focus on large scale trends, we also plot medians from each individual grouping of data points. The median value for each grouping is plotted with a thick circle. The error bar represents the interquartile range. Additionally, for more context, the bottom two panels in Fig. 3 show the dust optical depths and latitudes of the individual observations.

The sinusoidal curves shown in the top two panels of Fig. 3 have the form of Equa-216 tion 10, with the best-fit parameters A and ρ_o determined by fitting the grouped-median 217 values after removing observations during significant dust storms (where the dust op-218 tical depth is greater than 0.4). Note that the use of Equation 10 assumes the solar lon-219 gitudes of the minimum and maximum water levels are at 90° and 270° , respectively. The 220 fits were not used to determine the minimum and maximum locations because the large 221 data gaps prohibit a precise determination. Nonetheless, as discussed previously, the data 222 are consistent with these maximum and minimum values. 223

The best-fit parameters for the water density are $A = 115 \pm 90$ and $\rho_o = 1400 \pm$ 224 60. The best-fit parameters for the mixing ratio are $A = 1.3 \pm 0.7$ and $\rho_o = 11.4 \pm$ 225 0.42. The amplitudes, A, imply the water density and mixing ratio oscillate by $\sim 10\%$ 226 around their mean values over a Martian year. Our derived amplitude for the mixing ra-227 tio is smaller than the amplitude derived by Stone21 who found the oscillation around 228 the mean is $\sim 35\%$. Potential explanations for this discrepancy are the use of additional 229 data in our study, and the slightly different methodologies. The relatively small ampli-230 tudes also imply that variations on shorter timescales are quite significant and often more 231 prominent than the background seasonal trend. 232

It is interesting to compare the water levels at fixed seasons across different Mars 233 years. In general, year-to-year water levels are more consistent during the first half of 234 the year when atmospheric dust levels are lower and less variable. This can be seen in 235 Figure 3 by noting the median values are more tightly grouped around the best-fit curve 236 when L_s is less than 170°. Later in the year, when L_s is greater than 170°, the median 237 values are more spread around the best-fit curve. Near $L_s = 240^\circ$, the two median val-238 ues (MY33 and MY35) are highly separated around the best-fit curve, with the MY33 239 water levels being much higher than those in MY35. This large spread may be caused 240 by the different latitudes of the observations (MY33 are from the southern polar lati-241 tudes and MY35 are from the equatorial latitudes) or the atmospheric dust levels be-242 ing higher in MY33 than in MY35. Near $L_s = 340^\circ$, the median values are somewhat 243 244 separated around the best-fit curve and the water level is higher in MY35 than in MY32 and MY33. All three of these observational periods cover equatorial latitudes so a likely 245 explanation for the differences is the dust levels in the atmosphere, which were highest 246 during MY35. 247

Although there is a background seasonal variation in the thermospheric water abundance, the individual data points can deviate significantly from ta sinusoidal curve when observations from several years are combined. This suggest that the seasonal variation can change from year-to-year and that abrupt variations on shorter time scales are also significant. In the next section, we will explore these shorter term variations by exploring how water levels respond to changing atmospheric dust loading.

3.3 Dust Correlation

254

Figure 4 (top) shows the explicit correlation between the water mixing ratio and 255 the global averaged dust optical depth (Section 2). The averages of the log10 of the mix-256 ing ratios are plotted after grouping the data into dust optical depth bins of width 0.03. 257 The Pearson correlation coefficient between the bin-averaged mixing ratios and the bin-258 center dust optical depths is 0.95, indicating a strong correlation. From the lowest to high-259 est dust optical depths, the water mixing ratio increases from ~ 9 to ~ 60 (the highest 260 mixing ratios and dust optical depths are from the 2018 global dust storm (Montabone 261 et al., 2020)). Between these two extremes, the water mixing ratio steadily increases in 262 tandem with the dust optical depth, suggesting that atmospheric dust loading is a ma-263 jor driver of the thermospheric water cycle. This is consistent with the observed seasonal variation of the water mixing ratio, as the dust cycle peaks during southern spring and 265 summer when mixing ratios are highest. Further, because atmospheric dust levels dif-266 fer from year-to-year, one might expect the seasonal variation in the thermospheric wa-267 ter level to also change from year-to-year. 268

A linear fit to the log10 of the mixing ratios yields

$$log_{10}(\rho) = 1.4(\pm 0.08)X + 0.82(\pm 0.02) \tag{11}$$

where ρ is the mixing ratio and X is the dust optical depth. The best-fit slope, 1.4(±0.08), implies the relationship between mixing ratio and the dust optical depth can be approximated by a power law with a dependence that follows $\rho \propto 10^{1.4X}$.

The bottom two panels in Figure 4 show the water mixing ratio and local dust op-272 tical depth during two different time periods. Both of these periods from MY 35 are marked 273 with white boxes in the bottom panel of Figure 2 and coincide with the onset of a re-274 gional dust storm. During the first event in June 2020, MAVEN was crossing into the 275 northern hemisphere at low northern latitudes near 30°. The regional dust storm began 276 just after $L_s = 220^\circ$, when the dust optical depth began to rise. As the dust optical depth 277 increases, the water mixing ratio also increases. This dust storm was typical for this time 278 of year, demonstrating that the onset of typical regional dust storms leads to increased 279 thermospheric water. 280

During the second event (bottom panel) in November and December 2020, a "C-281 type" regional dust storm (Kass et al., 2016) began just after $L_s = 315^{\circ}$. MAVEN was 282 crossing back into the southern hemisphere near equatorial latitudes. At dust storm on-283 set, the water mixing ratio increases from ~ 15 to ~ 20 as the dust optical depth increases 284 from 0.25 to 0.35. At $L_s = 325^{\circ}$, the dust begins to dissipate and the mixing ratios de-285 crease back to pre-storm values. Nonetheless, during the onset of both dust events, the 286 water mixing ratio and dust optical depth are strongly correlated. However, each indi-287 vidual dust storm is somewhat unique and the response of thermospheric water might 288 vary depending on the location with respect to the dust storm.

²⁹⁰ 4 Discussion and Conclusions

MAVEN NGIMS observations allow us to infer water abundances and mixing ratios in the upper atmosphere of Mars to characterize their variability across more than three Martian years. At a fixed atmospheric pressure level of $\sim 5 \times 10^{-7}$ Pa, the typical water density is $1.3 \ (\pm 0.8) \times 10^3 \ \text{cm}^{-3}$ and the typical water mixing ratio is $10 \ (\pm 6)$ ppm. This average mixing ratio is a few times larger than that derived by Stone20, likely due to our somewhat different methodologies. Nonetheless, in agreement with Stone20, we find that water vapor is common in the upper atmosphere.

There is a seasonal variation in the thermospheric water, with higher water levels 298 during the second half of the year – which covers southern summer, perihelion, and dust 200 season $(L_s=180^{\circ}-L_s=360^{\circ})$ – and lower levels during the first half of the year during aphe-300 lion and non-dusty season. This is readily apparent when the water density and mixing 301 ratio is plotted as a time series (Figure 2). However, we find it difficult to quantify how 302 the water vapor varies as a function of solar longitude when the combining observations 303 from three different Mars Years (Figure 3). Instead, we find it is easier to quantify how 304 the water vapor varies as a function of the amount of dust in the lower atmosphere. This 305 leads us to conclude that the atmospheric dust level is a strong driver of the thermospheric 306 water. 307

In particular, we find the water vapor levels are consistent with a monotonically 308 increasing dependence on the global dust optical depth (Figure 4). This supports pre-309 vious work that has shown atmospheric heating from increased dust loading allows wa-310 ter to be transported upwards into the thermosphere without condensing (A. A. Fedorova 311 et al., 2020; M. S. Chaffin et al., 2021; Aoki et al., 2022). Our results suggest that this 312 process is the primary driver of thermospheric water vapor variability. Further support-313 ing this idea is that both regional and global dust storms are clear drivers of an increase 314 in thermospheric water . 315

Many questions remain about the variability of water vapor in the thermosphere such as, (1) how is the water vapor distributed with respect to local time and latitude?; (2) does the thermospheric water vapor depend on solar cycle?; and (3) how does the thermospheric water cycle affect the escape rate of hydrogen from Mars? These questions may be difficult to answer because there are limited observations after 2021 when MAVEN's orbit changed and its periapsis was raised to a higher altitude. Nonetheless, future modelling studies may use our results as constraints to explore these questions.

³²³ 5 Open Research

All of the MAVEN data used in this study are available at https://lasp.colorado .edu/maven/sdc/public/pages/datasets/ngims.html. The dust maps are available at http://www-mars.lmd.jussieu.fr/mars/dust_climatology/. The derived data products will be archived on zenodo.org upon aceptance of the manuscript.

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Figure 2. Derived thermospheric water abundances (top) and mixing ratios (middle). The bottom panel shows dust optical depth maps with the locations of the MAVEN observations shown in red. The dust optical depth data does not extend beyond early 2021. The two white boxes mark the regional dust storms that are analyzed in Section 3.3. The orange curves are sine functions that qualitatively demonstrate the seasonal dependence of water in the thermosphere.



Figure 3. Thermospheric water abundance and mixing ratio as a function of solar longitude (top two panels). The values from each orbit are shown with tiny circles, the grouped-median values are shown with large circles, and the error bars encompass the interquartile range. Colors mark the different Mars Years and the black curves are sinusoidal fits to the median values. The bottom two panels show the dust optical depth and latitude, respectively.



Figure 4. The top panel shows the thermospheric water mixing ratio as a function of globalaveraged column dust optical depth. Small circles show the individual data points, large circles show the average values after separating the data into dust optical depth bins. The black line is a best-fit to the binned averages. The middle panel and bottom panel each show the mixing ratio and dust optical depth during a regional dust storm in MY 35.