

# The Full Diurnal Cycle of Mars Water-Ice Cloud Optical Depth in EMIRS Observations

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## Key Points:

- Improved analysis of Mars thermal infrared spectra enabled retrieval of water ice cloud optical depth across the full diurnal cycle
- Observations showed higher water ice cloud abundance at night compared to day, with distinct peaks in early morning and evening hours
- Evening clouds were broadly distributed across low latitudes, while morning clouds were more concentrated over the Tharsis region

## 20 **Abstract**

21 Improvements to analyses of Martian thermal infrared spectra observed by the Emirates Mars  
22 Infrared Spectrometer (EMIRS) allow for retrieval of water-ice cloud optical depth,  $\tau_{\text{ice}}$ , across  
23 the full diurnal cycle at Mars. Using observations spanning nearly two Martian years, we find a  
24 persistent pattern of higher nighttime cloud abundance compared to daytime values, with regular  
25 peaks in  $\tau_{\text{ice}}$  occurring in both morning and evening hours. During the colder aphelion season  
26 when the low-latitude aphelion cloud belt forms, zonal mean optical depths ranged from maxima  
27 of  $\sim 0.5$  during the morning peak to minima of  $\sim 0.15$  near midday. Averaged across the full  
28 dataset, nighttime  $\tau_{\text{ice}}$  values were approximately twice the optical depth at the midday  
29 minimum. The morning peak in cloud abundance tended to occur near a solar incidence angle of  
30  $75^\circ$  across a range of latitudes and seasons. These diurnal patterns were generally consistent with  
31 modeled influences of atmospheric thermal tides on cloud formation conditions. Spatial  
32 differences in the distribution of clouds were noted across the diurnal cycle. Volcano clouds  
33 formed near midday and increased in optical depth throughout the afternoon and into the  
34 evening. At night, late evening clouds were observed more broadly across low-latitude regions,  
35 while morning clouds were more concentrated over the Tharsis region. The broad spatial and  
36 temporal coverage of these results enables new model comparisons against observations of  $\tau_{\text{ice}}$   
37 across the full diurnal cycle, and advances our understanding of water-ice cloud development  
38 and its role in the Mars climate system.

## 39 **Plain Language Summary**

40 Water-ice clouds are one of the most variable components of Mars' atmosphere. The abundance  
41 of these clouds changes across both daily and seasonal timescales, with notable variations  
42 occurring near surface features like volcanoes and large basins. Many previous observations  
43 were conducted near the middle of the day when these clouds are generally thinnest. Using new  
44 observations from the Emirates Mars Infrared Spectrometer (EMIRS) onboard the Emirates Mars  
45 Mission (EMM) spacecraft, we can now observe these clouds throughout the full Martian day  
46 and night. Looking at nearly two Martian years of observations, we found that clouds are  
47 typically thicker at night than during the day, with additional peaks in cloudiness occurring in the  
48 early morning and evening hours. Some of the thickest clouds form in a band near Mars' equator  
49 during the planet's cold season, reaching their daily maximum thickness shortly after sunrise.  
50 During this season, clouds in the late evening were found to be spread more widely across low  
51 latitudes, while early morning clouds were concentrated over a region called Tharsis. These  
52 results provide the first comprehensive observations of nighttime clouds on Mars, helping to  
53 validate atmospheric models and improve our understanding of how water moves through the  
54 Martian atmosphere.

## 55 **1 Introduction**

56 Water-ice clouds are among the most variable components of the Martian atmosphere over  
57 diurnal and spatial scales. Patterns of atmospheric temperature fluctuations driven by solar  
58 heating and thermal tides alter the environmental water saturation ratios that modulate cloud  
59 development (Zurek, 1976; Barnes et al., 2017; Clancy et al., 2017). This results in a noted  
60 coupling between the temperature profile and radiatively active water-ice clouds that vary  
61 throughout the full diurnal cycle of a Martian solar day (sol) (Madeleine et al., 2012b; Kleinböhl  
62 et al., 2013; Wilson & Guzewich, 2014; Szantai et al., 2021). At synoptic scales, variability in

63 cloud abundance has been linked with global circulation patterns that are affected by both  
64 dynamical processes and large-scale topography, in addition to the behaviors driven by diurnal  
65 and semi-diurnal thermal tides. At smaller scales, local topographic effects and small-scale  
66 circulations, such as those over volcanoes, can yield large spatial gradients in cloud abundance  
67 (Michaels et al., 2006; Wilson et al., 2007). In addition to these drivers of variability, water-ice  
68 clouds are themselves important factors controlling global circulation, radiative transfer, and the  
69 hydrologic cycle on Mars (Clancy et al., 2017; Montmessin et al., 2017; Määttänen &  
70 Montmessin, 2021). However, uncertainty remains over the temporal and spatial scales across  
71 which these clouds vary—particularly at night—leading to challenges in adequately modeling  
72 water-ice cloud and its effects in global circulation models (Määttänen & Montmessin, 2021),  
73 and in understanding the full role they play in the Mars planetary system.

74 Observations of variability in water-ice clouds began with terrestrial instruments and early  
75 spacecraft imagers and spectrometers on the Mariner and Viking missions, which provided the  
76 first indications of complexity across both spatial and temporal scales (e.g., Curran et al., 1973;  
77 French et al., 1981; Kahn, 1984). Quantification of water-ice cloud abundance across seasonal,  
78 inter-annual, and spatial dimensions was obtained from the Thermal Emission Spectrometer  
79 (TES) onboard the Mars Global Surveyor (MGS) orbiter, which produced the first detailed cloud  
80 climatologies as retrieved water-ice optical depth ( $\tau_{\text{ice}}$ ) (Smith, 2004). Numerous additional  
81 observational datasets have followed that expanded upon these results using instruments that  
82 observe across ultraviolet, optical, and infrared wavelengths (reviewed by Clancy et al., 2017  
83 and Määttänen & Montmessin, 2021)—yielding additional resolution and coverage, along with  
84 identification of new cloud morphologies, behaviors, and interactions. Among the findings of  
85 these results has been the identification of a seasonal water-ice cloud cycle that includes  
86 increases in cloud abundance during the colder aphelion season and decreases during the warmer  
87 and dustier perihelion season. Several cloud types have also been identified, with the most  
88 prevalent being those associated with the low-latitude aphelion cloud belt (ACB) and polar cap  
89 clouds that occur during the winter season in each hemisphere. However, many of these results  
90 were obtained from sun-synchronous instruments that observe at or near the same local true solar  
91 time (LTST) for all data points, leading to climatologies that are biased towards classification of  
92 afternoon behavior.

93 Information on the diurnal variability in water-ice clouds was generally more limited in early  
94 observational datasets, with some information on daytime variability produced from the Viking  
95 IRTM (Tamppari et al., 2003), and MGS Mars Orbiter Camera (Benson et al., 2003) instruments.  
96 More recent missions have begun to provide additional observations of daytime diurnal  
97 variability. Slowly migrating diurnal coverage from spacecraft such as Mars Express and Mars  
98 Odyssey have provided some information on daytime diurnal variability (e.g., Madeleine et al.,  
99 2012a; Smith, 2019)—though the slow orbital migration has led to the seasonal water-ice cloud  
100 signal being convolved with the diurnal signal. Additional diurnal observations have been  
101 provided at locations from surface missions (e.g., Kloos et al., 2018; Hayes et al., 2024; Smith et  
102 al., 2025). From these, a general conceptual picture of water-ice cloud daytime diurnal variability  
103 has emerged. In the case of clouds not associated with the seasonal polar cap clouds—between  
104 roughly 60 S and 60 N latitude, and occurring throughout the year but at greater magnitudes  
105 during aphelion season when they are associated with the ACB—water-ice clouds abundance  
106 generally reaches a minimum just after midday, with higher  $\tau_{\text{ice}}$  often observed in the mornings  
107 and afternoons (e.g., Smith, 2009, 2019; Pankine et al., 2013; Wolff et al., 2019, 2022; Giuranna  
108 et al., 2021; Szantai et al., 2021; Atwood et al., 2022, 2024). However, systematic diurnal

109 coverage necessary to validate and quantify this behavior has not been available until the  
110 Emirates Mars Mission (EMM)—which features observations from a spacecraft placed into a  
111 higher orbit suitable for systematic diurnal observations on sub-seasonal timescales.

112 Information about nighttime cloud abundance has been considerably more limited in most  
113 observational datasets, but has generally indicated higher water-ice optical depths during the  
114 night as compared to the day. Among the first indirect observational evidence for this behavior  
115 was data from studies that examined surface and atmospheric temperature anomalies at night.  
116 Wilson et al., (2007) used an atmospheric model with radiatively active water-ice clouds to  
117 attribute anomalously warm nighttime surface temperatures retrieved from some TES nighttime  
118 observations (~02:00 LTST) to increased downward infrared radiation from more abundant  
119 nighttime clouds. Similar indirect studies of coupling between atmospheric temperature profiles  
120 and water-ice cloud suggested higher nighttime cloud abundance would be expected given  
121 observed temperatures (Hinson & Wilson, 2004; Kleinböhl et al., 2013; Wilson & Guzewich,  
122 2014). Direct observations of  $\tau_{\text{ice}}$  at night have been obtained predominately from thermal  
123 infrared spectrometers, though such observations are less sensitive at night and subject to higher  
124 uncertainties (Pearl et al., 2001). Efforts to retrieve atmospheric opacities from TES nighttime  
125 observations have previously suggested higher  $\tau_{\text{ice}}$  at night as compared to the day (Pankine et  
126 al., 2013). Similar observations from the Planetary Fourier Spectrometer (PFS) onboard the Mars  
127 Express orbiter averaged into general “daytime” and “nighttime” bins showed similarly higher  
128 observed  $\tau_{\text{ice}}$  at night, along with some evidence for morning and afternoon peaks in abundance  
129 in ACB clouds, though with considerable variability still noted (Giuranna et al., 2021). Results  
130 from both TES and PFS showed differences in the spatial patterns of variability between day and  
131 night, but had insufficient diurnal and spatial coverage to capture the full behavior of nighttime  
132 clouds. Most recently, observations across the full diurnal cycle from the Thermal InfraRed  
133 Sensor (TIRS) on the Perseverance rover have found higher nighttime water-ice optical depths  
134 and a regular diurnal cycle to cloud behavior at Jezero crater through nearly two Mars Years of  
135 surface based measurements (Smith et al., 2025).

136 Modeling efforts have highlighted the importance of thermal tides in driving large-scale  
137 development of water-ice clouds, particularly at night, as well as the influence of radiatively  
138 active water-ice clouds on diurnal variability in the Mars surface-atmosphere system (e.g.,  
139 Hinson & Wilson, 2004; Madeleine et al., 2012b; Kleinböhl et al., 2013; Navarro et al., 2014;  
140 Pottier et al., 2017). Szantai et al., (2021) used the Mars Planetary Climate Model (PCM) to  
141 demonstrate diurnal patterns in water ice cloud abundance, correlating them with water ice cloud  
142 proxy measures from OMEGA spectra images. These models not only corroborated observed  
143 daytime variability but also predicted early morning peaks in water ice cloud abundance—  
144 attributing them to the influence of thermal tides—and the potential for greater cloud abundance  
145 at night. Despite these advancements, a comprehensive observational dataset spanning the full  
146 diurnal cycle of water-ice cloud necessary to validate many of these findings has been lacking  
147 (Määttänen & Montmessin, 2021).

148 In this study we focus on extending the observational record of variability in Mars water-ice  
149 cloud optical depth to include the first detailed coverage of the full diurnal cycle for low- and  
150 mid-latitude regions using thermal infrared observations from the Emirates Mars Infrared  
151 Spectrometer (EMIRS) onboard the Emirates Mars Mission (EMM) spacecraft. We first discuss  
152 the details of EMM observational coverage and the EMIRS dataset in section 2. This is followed  
153 by a discussion of the updated EMIRS retrieval algorithm and uncertainties in section 3,

154 including the implications and challenges associated with nighttime retrievals. Results are then  
155 presented in section 4 with a focus on the full diurnal cycle of  $\tau_{\text{ice}}$  shown across both spatial and  
156 seasonal scales for observations that span nearly two Martian years. In section 5 we compare the  
157 EMIRS results against other diurnal observations of water-ice cloud from the EXI imager that is  
158 also on-board EMM, and the TIRS instrument on the Perseverance rover. Finally, in section 6 we  
159 include a brief conclusion with a summary of the results and discussion of their potential use for  
160 future model comparisons. Data archive files that accompany this study include gridded products  
161 intended to better facilitate such comparisons.

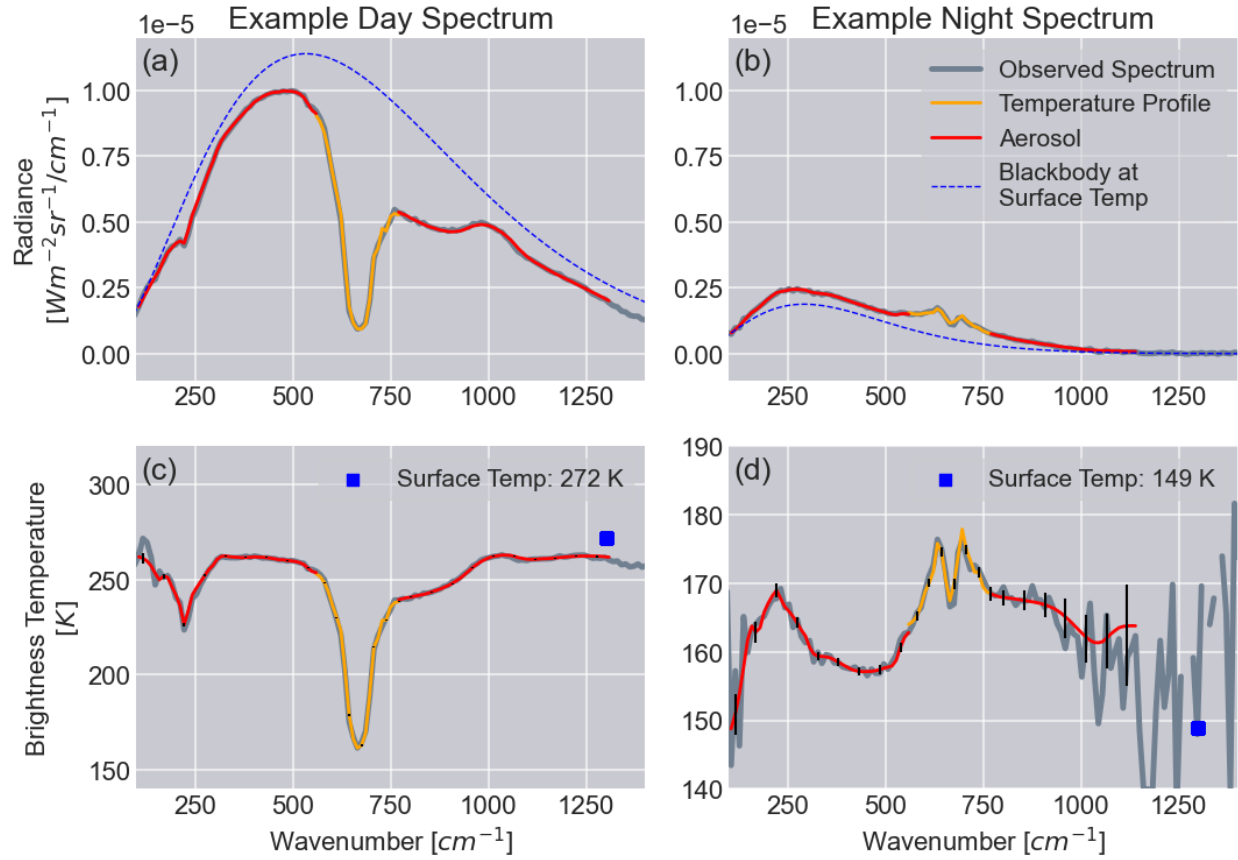
## 162 **2 EMIRS dataset**

163 EMIRS is a thermal infrared spectrometer onboard the Emirates Mars Mission (EMM) spacecraft  
164 (Almatroushi et al., 2021; Amiri et al., 2022). Observations produce spectra that include  
165 wavenumbers between approximately 100 and 1600  $\text{cm}^{-1}$  (wavelengths between 100 and 6  $\mu\text{m}$ )  
166 with a selectable spectral resolution of either 5 or 10  $\text{cm}^{-1}$  (Edwards et al., 2021). The  
167 instrument's projected observation footprint is elliptical, with the size of its major axis varying  
168 from approximately 100 to 300 km due to changes in spacecraft height and observation angle.  
169 The EMM spacecraft occupies a unique high-altitude, low-inclination elliptical orbit that varies  
170 between approximately 20,000 and 43,000 km. This orbital configuration allows for observations  
171 across all local times and nearly all latitudes and longitudes. EMIRS conducts raster scans of the  
172 Martian disk at various points in the 55-hour orbital period, typically achieving coverage across  
173 all local times and longitudes in roughly four orbits, or approximately  $5^\circ$  of solar longitude ( $L_s$ )  
174 on average. The result is a synoptic view of Mars with broad spatial and temporal coverage, and  
175 a resolution that allows for global and regional scale changes to be observed over periods of  
176 about two weeks. This allows for investigation of key aspects of diurnal variability in the  
177 Martian atmosphere, though shorter-timescale phenomena such as planetary waves may  
178 introduce variability that is not fully resolved in EMIRS observations. Standard retrieved  
179 properties include temperatures at the surface and for the atmospheric profile, optical depths for  
180 atmospheric dust and water-ice aerosol, and the column abundance of water vapor (Smith et al.,  
181 2022).

182 The dataset used in this study spans from the beginning of the EMM science phase at  $L_s \approx 49^\circ$  in  
183 Mars Year (MY) 36 through  $L_s \approx 347^\circ$  in MY 37 (approximately May 2021–October 2024). This  
184 period encompasses two aphelion seasons, allowing for inter-annual comparison of this  
185 climatologically significant period. Short gaps in the dataset exist due to spacecraft safe modes  
186 and a solar conjunction, during which no observations were made. Data from all local times were  
187 included in these analyses. All observations that passed initial quality control checks and had  
188 emission angles less than  $70^\circ$  were processed through the updated retrieval algorithm described  
189 further in the next section. Of these, approximately 2,040,000 retrievals yielded valid  $\tau_{\text{ice}}$  values  
190 and were included in subsequent analyses.

## 191 **3 EMIRS retrieval algorithm**

192 The retrieval algorithm is based on the constrained linear inversion approach originally  
193 developed for TES (Conrath et al., 2000; Smith, 2004; Smith et al., 2006), which retrieves  
194 atmospheric state parameters that best match each observed spectrum. Subsequent refinements  
195 were made to take advantage of the extended spectral range available in EMIRS observations  
196 and to improve treatment of surface thermophysical effects (Smith et al., 2022; Atwood et al.,



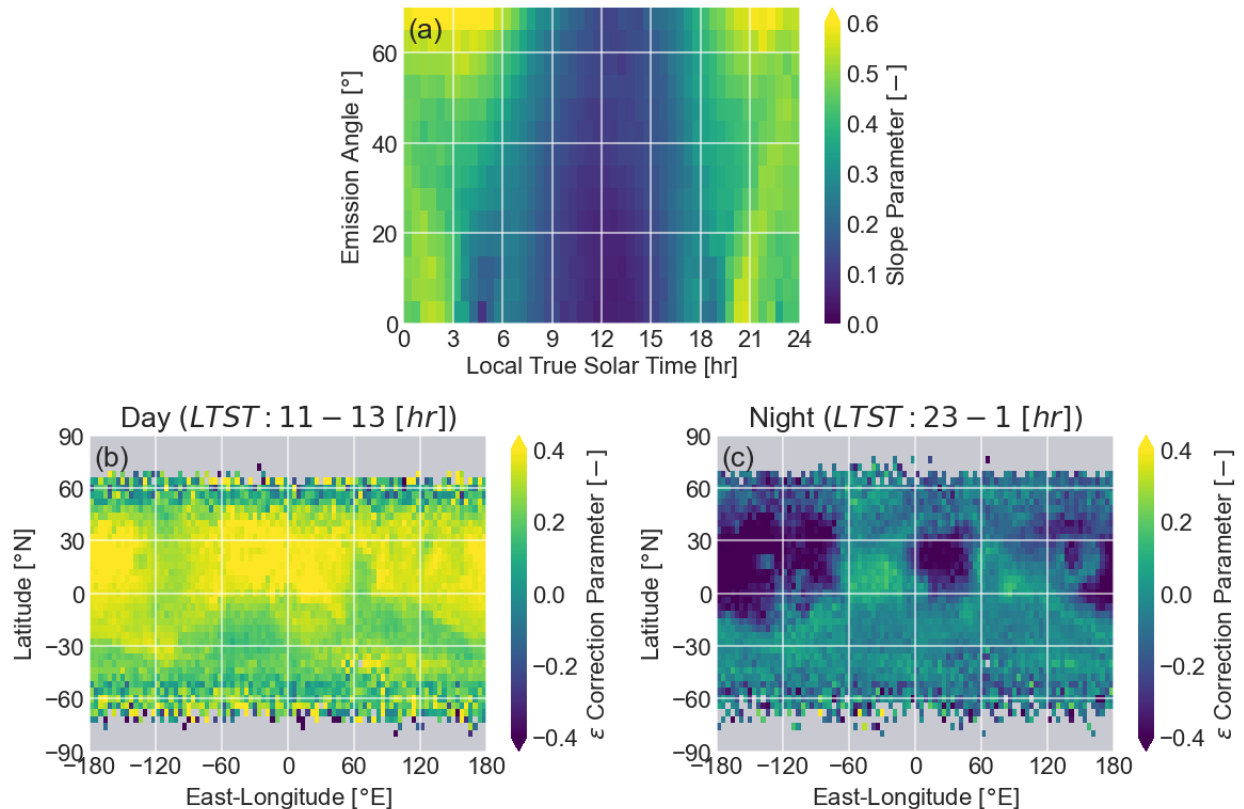
**Figure 1** Example EMIRS spectra for typical day (a,c) and night (b,d) observations with the presence of water-ice clouds. Observed spectra (grey lines) are shown as radiance (a,b) and equivalent brightness temperature (c,d). Orange and red lines show the forward model best-fit spectra for the regions of the spectrum used for the temperature profile and aerosol retrieval steps, respectively. The nominal surface temperature, retrieved at 1300 cm<sup>-1</sup>, is indicated with a blue box. A reference blackbody at this temperature is shown as the dashed blue line in the radiance plots. Black error bars in c&d show the wavenumber dependence of observation uncertainty as noise equivalent spectra radiance (NESR) converted to brightness temperature.

197 2024). The algorithm utilizes a forward radiative transfer model that includes multiple scattering  
 198 by aerosol particles to generate a spectrum that is then compared against observed EMIRS level  
 199 2 calibrated radiance spectra. Figure 1 illustrates example EMIRS spectra and associated best-fit  
 200 forward model results for both day and night observations as both radiance and equivalent  
 201 brightness temperature spectra. State parameters in the forward model are adjusted to achieve a  
 202 best fit to the observations (grey lines) via a two-step procedure. First, the atmospheric  
 203 temperature profile is retrieved from the CO<sub>2</sub> band centered at 667 cm<sup>-1</sup> (orange lines).  
 204 Subsequently, dust and water-ice optical depths are retrieved using portions of the spectrum  
 205 between 100 and 1315 cm<sup>-1</sup>, excluding the CO<sub>2</sub> band (red lines). Surface temperature is also  
 206 retrieved near 1300 cm<sup>-1</sup> during the second step (blue squares). For reference, radiance from a  
 207 blackbody at the retrieved surface temperature is shown in the radiance plots as a dashed blue  
 208 line.

209 Water-ice optical depth is retrieved as normal-incidence column-integrated full extinction values  
 210 referenced to  $825 \text{ cm}^{-1}$ . In the forward model, water-ice is distributed around an estimated  
 211 condensation layer (Smith, 2004; Atwood et al., 2022; Smith et al., 2022). However, given the  
 212 column integrated nature of the optical depth retrieval and large observation footprint relative to  
 213 typical cloud scales (e.g., Olsen et al., 2021; Patel et al., 2023), individual  $\tau_{\text{ice}}$  data points may  
 214 still include clouds at multiple layers or water-ice associated with more diffuse hazes distributed  
 215 across wider vertical regions. Previous work with this retrieval has indicated that uncertainty in  
 216  $\tau_{\text{ice}}$  associated with uncertainties in the vertical distribution of water-ice was generally less than  
 217 0.03 across seasonal average values (Atwood et al., 2022). For the results presented here, we  
 218 focus on retrieved  $\tau_{\text{ice}}$  and  $\tau_{\text{ice}}$  uncertainty, along with validation using additional parameters as  
 219 discussed below.

### 220 3.1 Surface thermophysical parameters

221 Surface thermophysical effects are incorporated into the forward model as spectrally dependent  
 222 emissivity properties drawn from lookup tables originally developed for TES (Bandfield, 2002).  
 223 Several additional empirical correction parameters are used in the forward model to correct for  
 224 expected deviations from these values. Surface anisothermality across the relatively large



**Figure 2** Mean values for surface thermophysical parameters used in the retrieval to correct for variability in observed spectra not directly accounted for in the forward model. (a) The linear slope parameter associated with variable surface temperature within an observation footprint. (b & c) The emissivity correction parameter during the day and night, respectively, showing similar patterns with known surface albedo and thermal inertia properties.

225 EMIRS observation footprint leads to slopes in the observed brightness temperature spectra  
226 (Bandfield & Edwards, 2008; Bandfield, 2009), which were parameterized via a linear slope  
227 parameter. An additional emissivity correction parameter was developed that applies to low  
228 wavenumbers where TES spectral data was limited and more uncertain due to a narrower  
229 spectral range and increased uncertainties below roughly  $250 \text{ cm}^{-1}$  as compared to EMIRS  
230 (Atwood et al., 2024). The resulting forward model radiance equation is discussed further in  
231 Smith (2004) and Smith et al. (2022), with updates that include details on the slope and  
232 emissivity parameters described in Atwood et al. (2024). Figure 2 illustrates the variability in  
233 these surface thermophysical parameters in retrievals for the EMIRS dataset.

234 Mean values for the slope parameter are shown as a function of LTST and emission angle in  
235 Figure 2(a). During midday, the diurnal cycle of surface heating from solar insolation led to  
236 more stable and homogeneous surface temperatures, resulting in lower values for this parameter  
237 (minimal spectral slope effect). At other local times, features within the observation footprint  
238 with different thermal inertias, albedos, or other properties exacerbated thermal inhomogeneities,  
239 resulting in larger values for the slope parameter. Higher emission angles, which yield both  
240 larger observation footprints and therefore potentially more regions with different temperatures,  
241 as well as increased spectral slopes when surface roughness or rock abundance are high  
242 (Nowicki & Christensen, 2007; Bandfield & Edwards, 2008; Bandfield, 2009), were also  
243 associated with larger slope parameter values.

244 The emissivity correction parameter is shown for midday (LTST 11:00–13:00) and midnight  
245 (LTST 23:00–01:00) mean values in Figures 2(b) and 2(c), respectively. During the day the  
246 pattern of variability showed similar characteristics to surface albedo, while during the night it  
247 more closely resembled that of thermal inertia (e.g., Jakosky et al., 2000; Nowicki &  
248 Christensen, 2007; Putzig & Mellon, 2007). While further investigation of the relationship of  
249 these empirical parameters to surface properties is beyond the scope of this work, we note that  
250 the results support classification of this variability as driven by thermophysical properties of the  
251 surface rather than unaccounted for atmospheric effects. Additional discussion and information  
252 about both of these parameters can be found in Atwood et al. (2024).

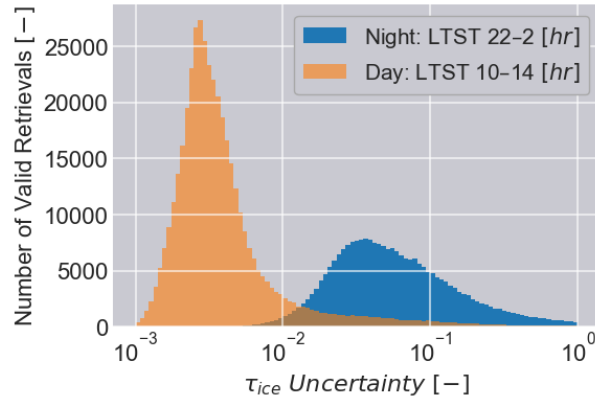


Figure 3 Histogram of water-ice optical depth uncertainty for all valid retrievals during daytime (LTST = 10:00 – 14:00 [hr]) and nighttime (LTST = 22:00 – 2:00 [hr]) observations.

### 253 3.2 Nighttime retrievals

254 Nighttime retrievals of thermal infrared spectra present unique challenges due to lower observed  
 255 radiance and reduced thermal contrast between the surface and atmosphere (illustrated in Figures  
 256 1b&d, respectively). These factors result in decreased sensitivity to water-ice cloud properties  
 257 and increased uncertainty in the  $\tau_{ice}$  values. Primary sources of uncertainty are related to the  
 258 observation signal-to-noise ratio and thermal contrast, with additional important factors related to  
 259 surface thermophysical effects (e.g. thermal inertia and albedo, among others), errors in cloud  
 260 vertical distributions, and variability in water ice particle size distributions (discussed further in  
 261 Edwards et al., 2021; Smith et al., 2022; Atwood et al., 2024). As part of the retrieval algorithm,  
 262 uncertainty in retrieved  $\tau_{ice}$  is estimated using numerical sensitivity tests following the method  
 263 outlined in Smith et al. (2022). This produced an estimate that accounts for both signal-to-noise  
 264 ratios that vary greatly between day and night observations, and changes to thermal contrast that  
 265 affect the ability of thermal infrared observations to differentiate between components of the  
 266 surface and atmosphere. In daytime observations,  $\tau_{ice}$  uncertainties were typically below 0.05,  
 267 and often below 0.01 for observations outside of polar cap regions. Within these results,  $\tau_{ice}$   
 268 values for individual spectra ranged from below 0.05, indicating minimal water-ice cloud, to  
 269 values above 1 for the largest observed abundances.

270 The inclusion of the updated surface thermophysical parameters and the 45- $\mu\text{m}$  water-ice cloud  
 271 band into the EMIRS retrieval considerably improved the performance of the  $\tau_{ice}$  retrievals at  
 272 night. Nevertheless,  $\tau_{ice}$  uncertainties ranged from less 0.001 to greater than 1 in these results  
 273 (Figure 3). To accommodate the increased uncertainty while maintaining a comprehensive  
 274 dataset, quality control criteria within the algorithm have been adjusted and analyses have been  
 275 updated to better account for this range of  $\tau_{ice}$  uncertainty values. Previous analyses focused on  
 276 daytime observations using this retrieval algorithm have limited results to spectra with  $\tau_{ice}$   
 277 uncertainties below 0.05 (e.g., Smith, 2004; Smith et al., 2022; Atwood et al., 2022). In this  
 278 study, we utilized a  $\tau_{ice}$  uncertainty threshold of 1.0, considering all retrievals below this value  
 279 valid if they passed other quality control checks. This threshold value was selected to ensure  
 280 inclusion of typical nighttime data (Figure 3), while still removing data points with very high  
 281 uncertainties that may represent invalid data points or failures of the retrieval algorithm.

282 Additionally, as optical depths less than zero are not possible in the retrieval algorithm, when  
 283 uncertainties are of the same scale as the optical depth values (which is particularly common at  
 284 night) a potential bias in mean optical depth estimates can be introduced. To account for this in  
 285 our analyses we employed an updated uncertainty-weighted averaging method that de-weights  
 286 retrieved values with higher  $\tau_{ice}$  uncertainty when averaging across various temporal and spatial  
 287 dimensions. As the range of  $\tau_{ice}$  values within each averaging grid box would be expected to be a  
 288 function of both natural variability between observations and uncertainty in the observation  
 289 itself, the weights applied to each retrieved spectrum are estimated as a function of both metrics.

290 The natural variance of  $\tau_{ice}$  within a grid box,  $j$ , was estimated as the mean squared error,

$$291 \quad v_{nat,j} = \frac{1}{N} \sum_i^N (\tau_{ice,i} - \overline{\tau_{ice}})^2 \quad (1)$$

292 for each retrieval,  $i$ , of  $N$  total retrievals with a mean value,  $\overline{\tau_{ice}}$ . The expected total variance was  
 293 then calculated by sums of variance for both natural variability and uncertainty,  $\sigma_i$ , to produce a  
 294 weight for each spectrum as,

$$295 \quad w_i = \frac{1}{v_{nat,j} + \sigma_i^2} \quad (2)$$

296 This estimate relies on the assumption that natural variability and uncertainty are uncorrelated,  
 297 and that variability due to uncertainty therefore does not bias the  $v_{nat,j}$  estimate within the grid  
 298 box. A simple investigation of this assumption was conducted through calculation of the Pearson  
 299 correlation coefficient between the  $\tau_{ice}$  anomaly (i.e.,  $\tau_{ice,i} - \overline{\tau_{ice}}$ ) and the uncertainty for each  
 300 datapoint. This coefficient was calculated for each of the various averaging dimensions shown  
 301 below in the results sections. The correlation coefficients ranged between 0.24 and 0.32,  
 302 indicating a small positive correlation between the two. However, the associated coefficients of  
 303 determination ( $R^2$ ) were less than 0.1—indicating that for a simple linear correlation model, no  
 304 more than  $\sim 10\%$  of the variability in  $v_{nat,j}$  could be explained by  $\sigma_i^2$ , or vice versa. For the  
 305 purposes of this weighted averaging, this was considered sufficiently independent to apply the  
 306 simple sums of variance method to estimate a weighting factor. Finally, the weighted mean of  
 307  $\tau_{ice}$  for each grid box was estimated as,

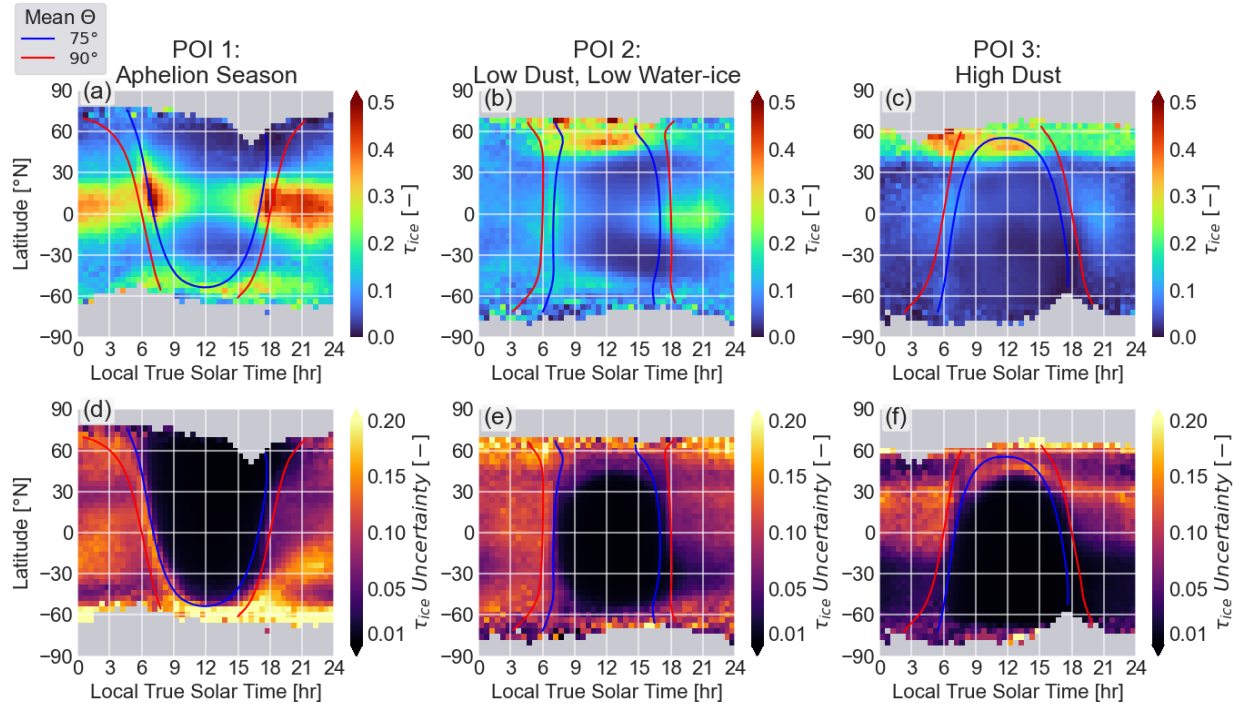
$$308 \quad \tau_{ice,j} = \frac{\sum_i^N w_i * \tau_{ice,i}}{\sum_i^N w_i} \quad (3)$$

309 Two alternative averaging methods using an un-weighted mean and a weighted mean dependent  
 310 only on the uncertainty (i.e.,  $w_i = 1/\sigma_i^2$ ) were also calculated for reference. During low-  
 311 uncertainty periods during the day the differences between averaging methods were minimal,  
 312 while differences increased at night and when uncertainties were high. Across all valid grid  
 313 boxes with more than 10 data points used in the subsequent sections, and in comparison to the  
 314 updated (uncertainty and natural variability weighted) method, the un-weighted mean was  
 315 approximately 13% higher on average (standard deviation of the grid box percentage differences:  
 316 29%) while the uncertainty-only weighted mean was approximately 17% lower (standard  
 317 deviation: 28%). Results in the following sections utilize this updated weighted-mean  
 318 methodology using equations 1–3 for all  $\tau_{ice}$  analyses. In addition, mean  $\tau_{ice}$  uncertainty is shown

319 for each figure to illustrate changes in water-ice sensitivity across various dimensions and place  
320 typical uncertainty values in context.

#### 321 **4 Results and discussion**

322 Results were first categorized into three seasonal periods of interest (referred to here as POI)  
323 based on typical water-ice cloud and dust conditions. The first of these was classified as a simple  
324 aphelion season average (POI 1), considered here as  $L_s=40^\circ-150^\circ$  in MY 36 & 37, during which  
325 the ACB was apparent and low-latitude water-ice clouds were abundant. For observations  
326 outside this  $L_s$  range, two additional categories of spectra were classified based on the amount of  
327 dust in the atmosphere (see Smith et al., 2025a, for additional discussion of  $\tau_{\text{dust}}$  in these results).  
328 For periods during and just after regional dust storms, the elevated concentration of atmospheric  
329 dust warms the atmosphere and suppresses water-ice cloud formation. These periods of higher  
330 dust concentrations (POI 3) were classified as: MY 36  $L_s=220^\circ-270^\circ$ , MY 36  $L_s=312^\circ-340^\circ$ ,  
331 MY 37  $L_s=210^\circ-270^\circ$ , and MY 37  $L_s=310^\circ-325^\circ$ . The final category (POI 2) included all  
332 remaining periods, and was generally classified as having lower water-ice concentrations as  
333 compared to POI 1 and lower dust concentrations as compared to POI 3.



**Figure 4** (a–c) EMIRS zonal mean water-ice optical depth across the full diurnal cycle for observations during (left column) POI 1: aphelion season ( $L_s=40^\circ\text{--}150^\circ$ ), (middle column) POI 2: periods between regional dust storms, and (right column) POI 3: dusty periods during and after regional dust storms, with (d–f) associated mean uncertainties. Contour lines are shown for the mean solar incidence angle ( $\theta$ ) at  $90^\circ$  (red; the terminator between day and night) and  $75^\circ$  (blue).

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#### 4.1 Zonally averaged diurnal cycle

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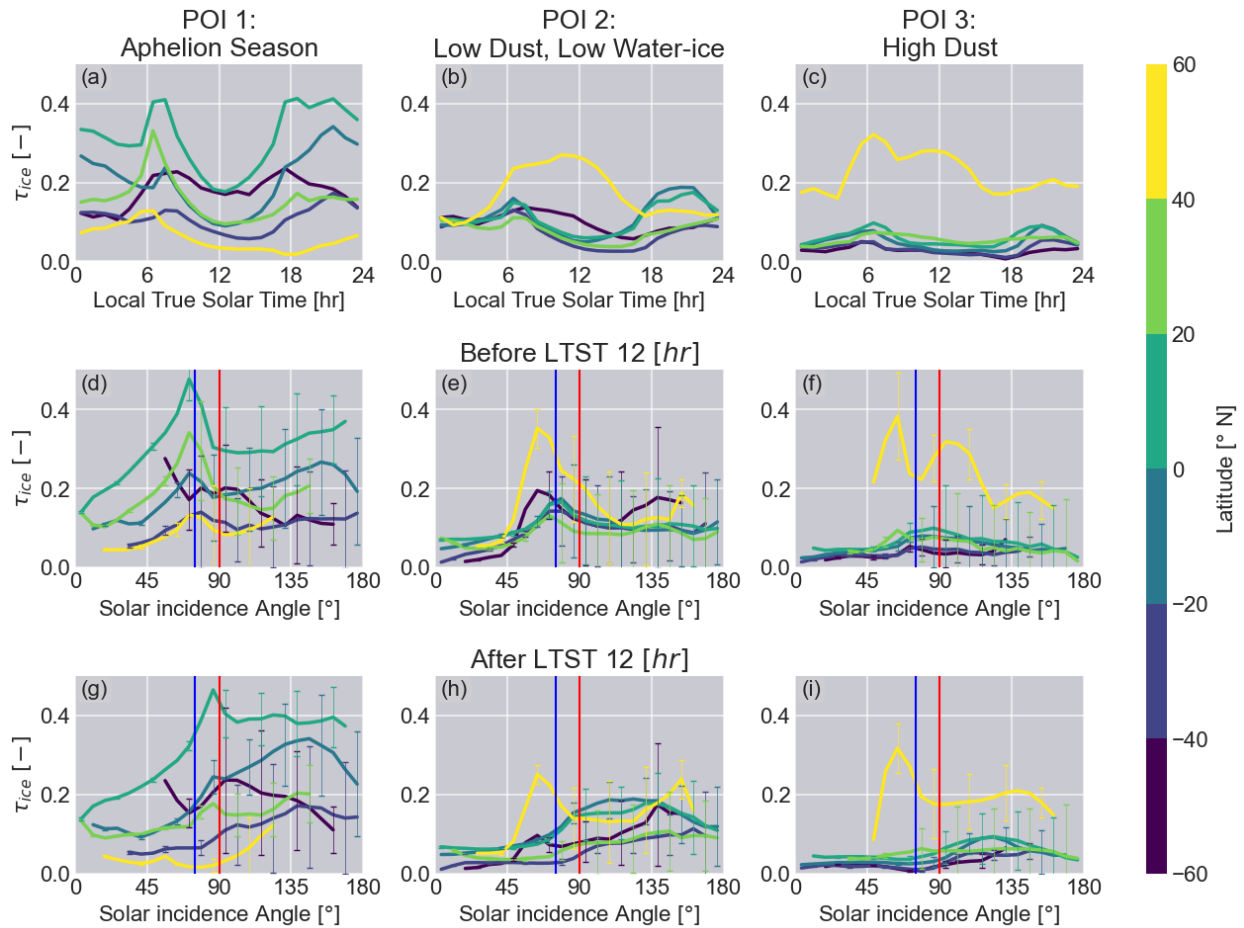
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The full diurnal cycle of retrieved water-ice optical depth is shown as zonally averaged  $\tau_{ice}$  values for the three periods of interest in Figure 4. During the aphelion season average (Figure 4a), a strong diurnal cycle was evident in the low-latitude ACB. The midday minimum in  $\tau_{ice}$  noted in previous datasets was similarly apparent in these results. Cloud abundance began to increase from this midday minimum, including a rapid increase and small peak near the evening terminator, followed by a decreasing trend throughout the night. However, as will be discussed further below, longitudinal differences indicated the timing of this nighttime maximum in  $\tau_{ice}$  varied by region. Shortly after dawn  $\tau_{ice}$  again increased to the largest zonal-mean values in the diurnal cycle. Nighttime clouds were similarly highest in the low-latitudes of the ACB, but were observed with wider latitudinal extents than the daytime ACB. The daytime ACB has often been associated with clouds between roughly  $10^\circ\text{S}$  and  $30^\circ\text{N}$  latitude (e.g., Clancy et al., 2017). The nighttime ACB was somewhat wider, with equivalent  $\tau_{ice}$  threshold values expanding the ACB to regions between roughly  $25^\circ\text{S}$  and  $35^\circ\text{N}$  at its maximum extent around 22:00.

The average solar incidence angle is shown in these plots for  $90^\circ$  (red contour line) indicating the terminator at dawn and dusk, and at  $75^\circ$  (blue contour line), which indicates times shortly after dawn and before dusk ( $15^\circ$  is one Martian hour at the equator on the equinox). Interestingly, the

351 ACB morning  $\tau_{ice}$  peak closely followed the  $75^\circ$  solar-incidence angle contour across a range of  
 352 latitudes (Figure 4a). Zonal average  $\tau_{ice}$  for  $20^\circ$  latitude bins are shown in Figure 5 as a function  
 353 of LTST (top row), and solar incidence angle for the dawn and dusk segments (middle and  
 354 bottom row, respectively). During aphelion season the local time of this early morning peak  
 355 ranged by latitude from roughly 06:00 to 08:00 (Figure 5a) but was consistently near  $75^\circ$  solar-  
 356 incidence angle (except for polar cap clouds at higher southern latitudes), indicating that the peak  
 357 may be related to time since sunrise. While there was a peak in  $\tau_{ice}$  during the evening for low- to  
 358 mid-latitude clouds in these results, there was no evidence for a consistent relationship in the  
 359 timing of that peak to either local time or solar incidence angle. Rather, the late  $\tau_{ice}$  peak tended  
 360 to occur between roughly sunset and midnight. Similar diurnal patterns tended to occur in POI 2  
 361 during periods between dust storms throughout the rest of the year, though at lower magnitudes.  
 362 During dust storms water-ice cloud abundance was generally quite low, though a similar pattern  
 363 of relative peaks in the morning and evening were present. Across the full dataset, zonal-mean



**Figure 5** (a–c) EMIRS zonal mean water-ice optical depths across the full diurnal cycle for similar observation periods to Fig 3. Results are further segmented by solar incidence angle for observations before (d–f) and after (g–i) local noon. Uncertainties are shown as error bars at  $\pm$ -mean uncertainty (included only on d–i to improve figure legibility) and show the generally higher uncertainties typical of nighttime (solar incidence angles  $> 90^\circ$ ) observations. Vertical lines show equivalent solar incidence angles to Fig 3 at  $75^\circ$  and  $90^\circ$ .

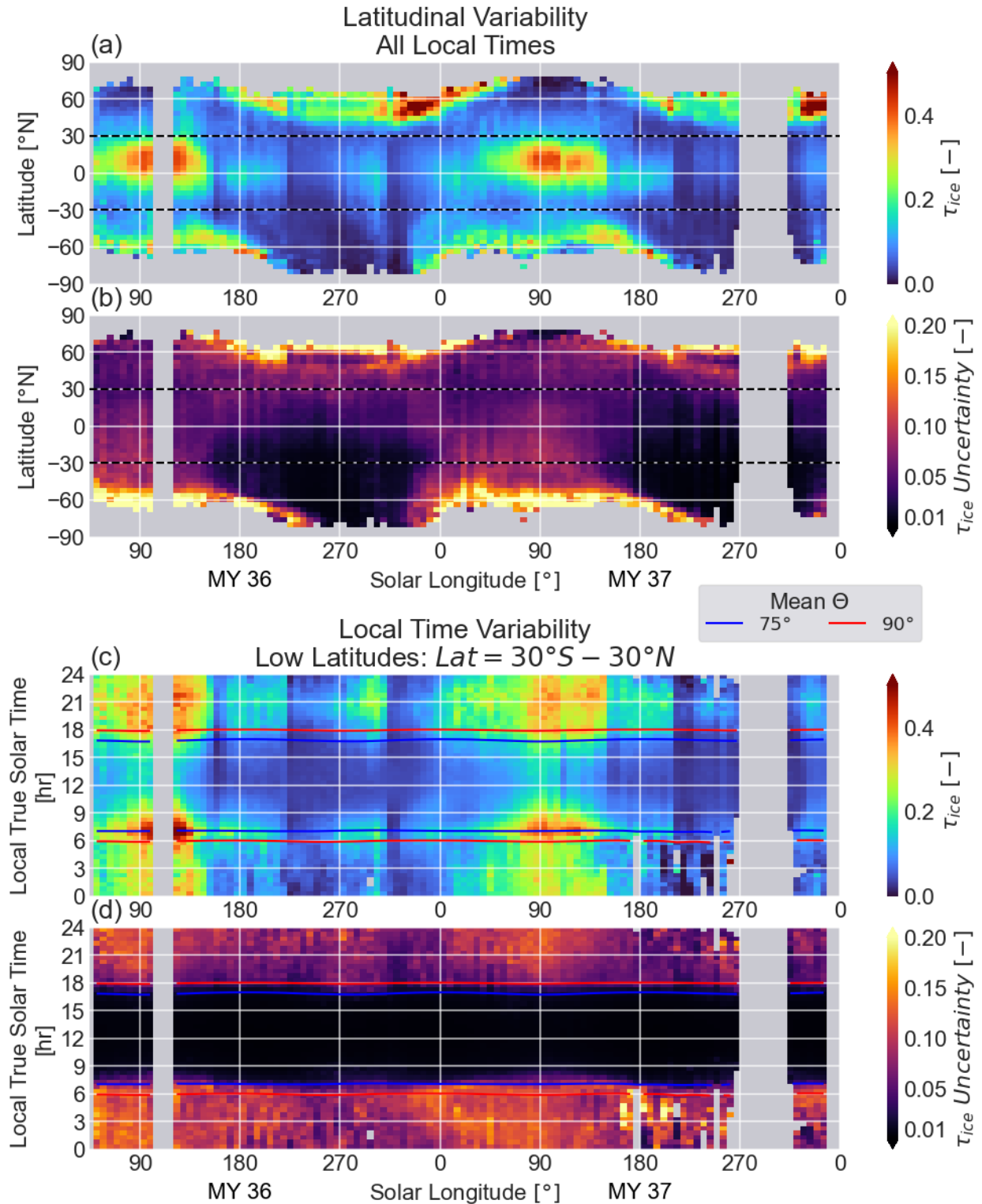
364  $\tau_{\text{ice}}$  values observed at night were approximately twice their midday minimum values. This  
365 nighttime enhancement was most pronounced in the ACB where midday values of  $\sim 0.15$  were  
366 contrasted with nighttime values that often exceeded 0.3.

367 Mean  $\tau_{\text{ice}}$  uncertainty patterns shown in Figure 4 tended to follow the diurnal cycle, with  
368 generally higher uncertainties during nighttime hours. As uncertainty is primarily driven by total  
369 radiance and thermal contrast between the atmosphere and surface, some additional complexity  
370 was apparent in the patterns. A peak in uncertainty occurred just after dawn (red contour lines)  
371 during aphelion season as solar heating began warming the surface. During these times, surface  
372 temperatures tended to transition from below atmospheric temperatures to above them, resulting  
373 in reduced thermal contrast and decreased sensitivity to water-ice spectral features. Notably, the  
374 observed morning peak in  $\tau_{\text{ice}}$  occurred after this uncertainty maximum, suggesting it represents  
375 a real feature rather than an artifact of retrieval uncertainty. The error bars in Figure 5, which  
376 show mean uncertainty around the zonal mean optical depth, provide additional context for  
377 comparing uncertainty magnitudes to diurnal  $\tau_{\text{ice}}$  variations. For low-latitude ACB clouds in  
378 particular, while nighttime uncertainties were higher than during the day, the magnitude of the  
379  $\tau_{\text{ice}}$  variability was greater than nighttime uncertainties, supporting a robust finding of higher  
380 mean water-ice cloud abundance at night compared to day during aphelion season.

381 Finally, we also note that some degree of correlation between  $\tau_{\text{ice}}$  and  $\tau_{\text{ice}}$  uncertainty was  
382 apparent in these results, particularly at night. As noted in section 3.2, uncertainties of the same  
383 scale as optical depth (combined with a non-negative constraint on optical depth) would be  
384 expected to bias mean  $\tau_{\text{ice}}$  values high. While some of this effect was removed by the updated  
385 uncertainty weighted mean method, we note that a “background” component of the nighttime  
386 optical depth signal related to increased uncertainties may still exist in these results.  
387 Nevertheless, the noted independence of major optical depth features from uncertainty features  
388 in figure 4 supports the general conclusions of these results.

#### 389 4.2 Seasonal trends in the diurnal cycle

390 The seasonal progression of water-ice cloud optical depth is shown in Figure 6 for observations  
391 spanning from  $L_s \approx 49^\circ$  in MY 36 through  $L_s \approx 347^\circ$  in MY 37. Zonal mean  $\tau_{\text{ice}}$  averaged across  
392 all local times (Figure 6a) showed patterns consistent with previous observations of the seasonal  
393 water-ice cloud cycle (e.g. Smith, 2004). Polar cap clouds were apparent in winter hemispheres,  
394 while a seasonal cycle in low- and mid-latitude clouds reached a peak during aphelion seasons  
395 ( $L_s \approx 40^\circ$ - $150^\circ$ ) associated with ACB clouds. However, mean  $\tau_{\text{ice}}$  values in EMIRS results were  
396 higher than those reported from TES observations—which were obtained at approximately 14:00  
397 LTST near when water-ice clouds typically reach their daily minimum. While TES maximum  
398 values for ACB clouds were around 0.15, the inclusion of all local times in these averages  
399 yielded EMIRS mean  $\tau_{\text{ice}}$  values between 0.4 and 0.5 during peak ACB periods—reflecting the  
400 substantial diurnal variability and increased nighttime abundance of water-ice clouds.  
401 Additionally, EMIRS observations revealed increased cloud abundance during less dusty periods  
402 outside of aphelion season. These clouds, which formed primarily at night and were often  
403 associated with distinct evening and morning peaks, were not readily apparent in the midday  
404 TES observations.



**Figure 6** The seasonal progression of EMIRS water-ice optical depth and associated uncertainty, between roughly  $L_s=40$  in MY 36 to  $L_s=347$  in MY 37. (a,b) Zonal mean optical depth by latitude averaged across all local times. (c,d) The diurnal cycle averaged across low latitudes ( $30^{\circ}S - 30^{\circ}N$ ) and all longitudes. Contour lines shown similar to Figure 3.

406 The diurnal cycle of  $\tau_{\text{ice}}$  averaged across low latitudes ( $30^{\circ}\text{S} - 30^{\circ}\text{N}$ ) is shown in Figure 6c. The  
 407 fundamental pattern noted in the previous section—characterized by morning and evening peaks  
 408 separated by a midday minimum and often higher opacities at night—persisted throughout much  
 409 of the Martian year, though with considerable variation in magnitude. During aphelion seasons,  
 410 low-latitude ACB clouds reached zonal mean  $\tau_{\text{ice}}$  values as high as 0.5 in the early morning  
 411 while dropping to  $\sim 0.15$  near midday. This diurnal pattern remained detectable, but with reduced  
 412 amplitude and some changes in the relative amplitudes of the morning and evening peaks,  
 413 throughout much of the rest of the year. Non-aphelion season low dust periods included  $\tau_{\text{ice}}$   
 414 values that exceeded 0.2, as well as some periods in which the diurnal maximum in  $\tau_{\text{ice}}$  occurred  
 415 during the evening peak. During regional dust storms, increased atmospheric dust loading and  
 416 the associated atmospheric warming tended to suppress water-ice cloud formation at night and  
 417 limit the morning and evening peaks in water-ice optical depth.

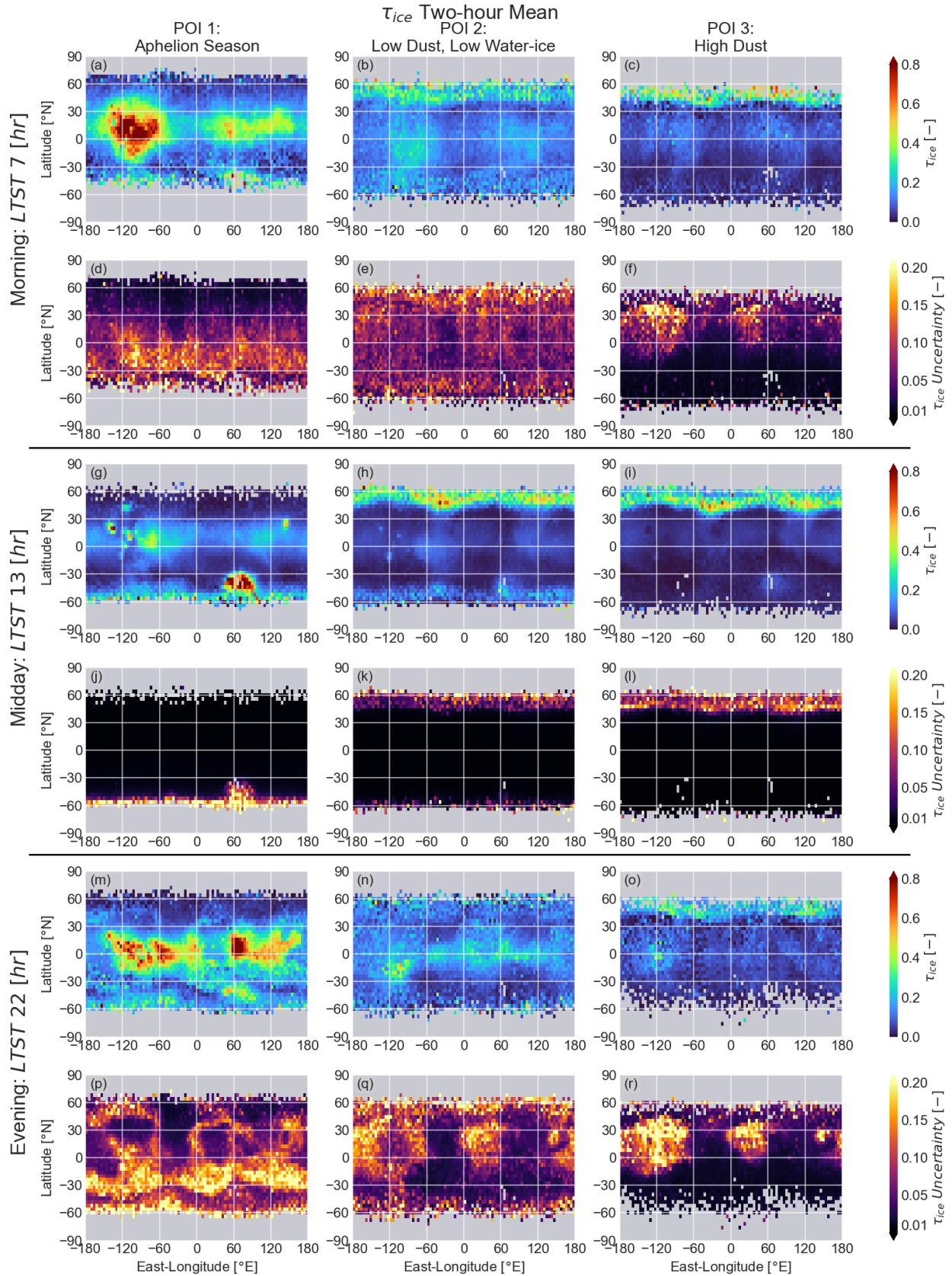
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419 Mean uncertainties associated with these retrievals showed expected patterns of higher values  
 420 during nighttime hours (Figures 6d). Nevertheless, the seasonal and diurnal patterns in  $\tau_{\text{ice}}$   
 421 remained robust against these uncertainty ranges, particularly during aphelion seasons when  
 422 water-ice clouds were most abundant. Even during non-aphelion periods of elevated water-ice  
 423 abundance, differences between morning and evening peaks in  $\tau_{\text{ice}}$  and midday minima were  
 424 often greater than mean nighttime uncertainties.

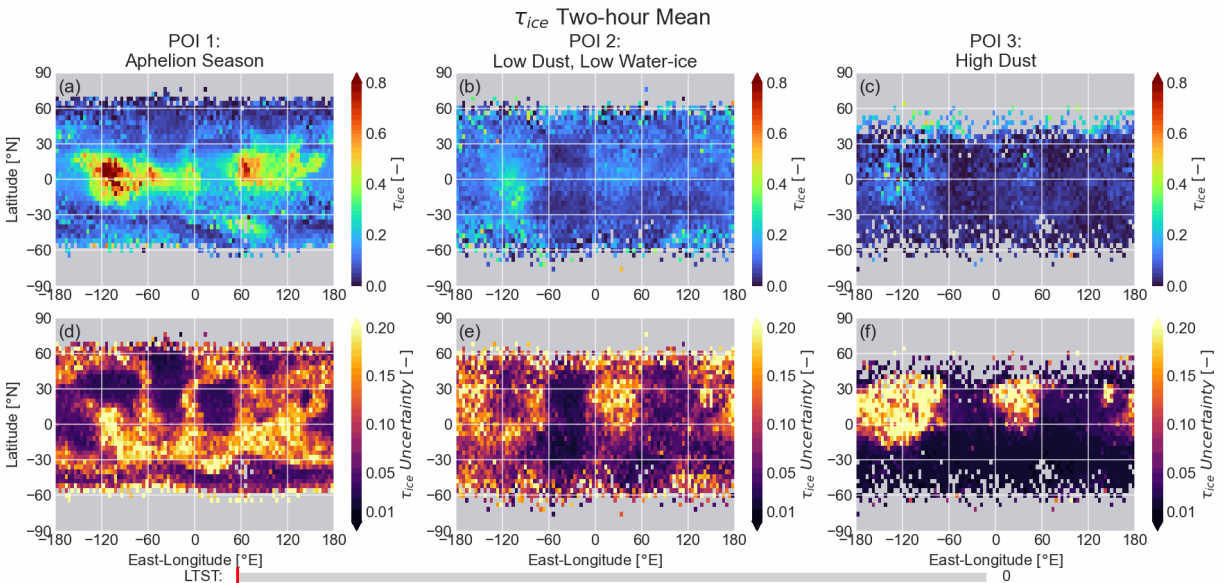
#### 425 4.3 Spatial variability

426 Spatial maps of uncertainty-weighted mean  $\tau_{\text{ice}}$  for 2-hour rolling-average periods during  
 427 morning (centered at 07:00 LTST), midday (13:00 LTST), and late evening (22:00 LTST) are  
 428 shown in Figure 7 for the three periods discussed previously. An animation showing these maps  
 429 across all hours of the diurnal cycle is available online as Figure 8. During the daytime,  
 430 development of water-ice clouds over volcanoes followed previously reported patterns (e.g.,  
 431 Atwood et al., 2022; Wolff et al., 2022), with formation beginning near midday and clouds  
 432 growing in both spatial extent and optical depth throughout the afternoon. This pattern was  
 433 evident not only during aphelion season but also during less dusty periods between regional dust  
 434 storms, when  $\tau_{\text{ice}}$  values often exceeded 0.7 in the vicinity of volcanoes.

435 Evening peaks in water-ice abundance showed considerable spatial variability, with  $\tau_{\text{ice}}$   
 436 exceeding 0.7 across a range of longitudes and regions. The timing of this peak varied by region  
 437 as well, with the Tharsis and volcano evening peaks observed closer to 18:00 LTST, while other  
 438 regions (e.g.  $\sim 75^{\circ}\text{E}$  longitude over Syrtis Major) included an evening peak in  $\tau_{\text{ice}}$  closer to 22:00  
 439 LTST. This broad distribution of water-ice clouds in the evening was contrasted with the early  
 440 morning hours between roughly 03:00 LTST and the morning peak around  $\sim 07:00$  LTST—when  
 441 the highest optical depths (exceeding 0.7) were concentrated over the Tharsis region, while other  
 442 low-latitude clouds between roughly  $30^{\circ}\text{E} - 150^{\circ}\text{E}$  longitude showed lower values around 0.4.  
 443 During aphelion season, these early morning Tharsis ACB clouds extended from roughly  $25^{\circ}\text{S}$  to  
 444  $45^{\circ}\text{N}$ , indicating the substantial longitudinal enhancement of the morning ACB was largely due  
 445 to clouds in the Tharsis region.



**Figure 7** Spatial maps of 2-hr uncertainty-weighted mean EMIRS water-ice optical depth and mean uncertainty for the three periods of interest shown in Fig 3 and discussed in the text. Time periods shown here are for the morning peak in optical depth (diurnal maximum over Tharsis) at LTST=07:00 (top), midday diurnal minimum at LTST=13:00 (middle), and in the late evening at LTST=22:00 (bottom). Rolling 2-hr averages throughout the full diurnal cycle are shown as an animation in the online version in Fig 8.



**Figure 8** Spatial maps of 2-hr uncertainty-weighted mean EMIRS water-ice optical depth and mean uncertainty for the three periods of interest shown in Fig 3 and discussed in the text. Rolling 2-hr averages throughout the full diurnal cycle are shown as an animated version of Figure 7.

447 Mean uncertainties associated with nighttime retrievals revealed patterns related to surface  
 448 thermophysical properties. As discussed in section 3.1, variable thermal inertia across different  
 449 surface types lead to variability in  $\tau_{ice}$  uncertainties at night. Higher uncertainties tended to occur  
 450 in regions and at local times where decreased thermal contrast between the surface and  
 451 atmosphere reduced sensitivity to water-ice spectral features.

#### 452 4.4 Drivers of diurnal variability in $\tau_{ice}$

453 Finally, the observed pattern of increased water ice abundance at night and peaks in  $\tau_{ice}$  during  
 454 morning and evening hours showed similarities to previous studies that have suggested such  
 455 patterns may be due in part to thermal tide and topographic effects (e.g., Hinson & Wilson, 2004;  
 456 Madeleine et al., 2012b; Kleinböhl et al., 2013; Wilson & Guzewich, 2014; Pottier et al., 2017;  
 457 Szantai et al., 2021). Previous modeling work has indicated thermal tides as a likely driver of  
 458 diurnal variability in water ice clouds, particularly through their influence on atmospheric  
 459 temperatures that modulate conditions for cloud development. Modeling results from Hinson &  
 460 Wilson (2004) found spatial patterns in  $\tau_{ice}$  at 14:00 and 04:00 that were largely consistent with  
 461 the EMIRS observations at these times, and attributed much of the differences in water-ice  
 462 opacity between the two time periods to the effect of thermal tides. These studies have

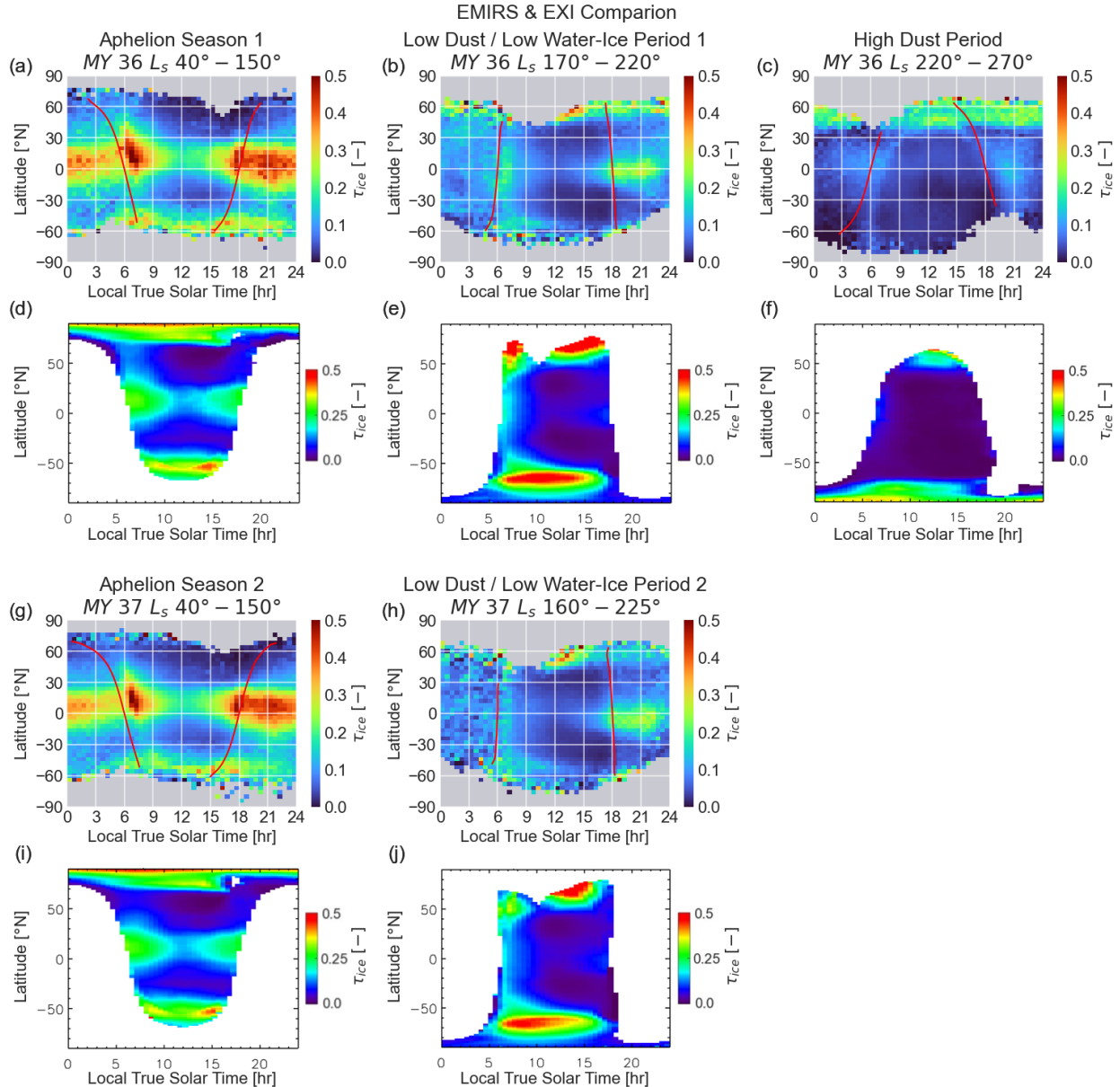
463 highlighted feedback between radiatively active water-ice clouds, thermal tides, and topography  
464 as key potential drivers of diurnal variability in water-ice cloud abundance. While a detailed  
465 investigation of this relationship is beyond the scope of this work, we note that the spatial and  
466 temporal coverage in EMIRS results may allow for further investigation or validation of these  
467 feedbacks in models of the Martian atmosphere.

## 468 **5 Comparison with other diurnal $\tau_{\text{ice}}$ observations**

469 The Thermal InfraRed Sensor (TIRS) is a set of upward-looking radiometers onboard the  
470 Perseverance rover that measures downwelling thermal infrared radiation, enabling retrieval of  
471 aerosol optical depth at the low-latitude Jezero crater (18.4°N, 77.5°E) across the full diurnal  
472 cycle (Rodriguez-Manfredi et al., 2021; Smith et al., 2023). Recent analysis of nearly two Mars  
473 Years of TIRS observations (over a similar time period to the EMIRS dataset) has enabled  
474 separation of dust and water-ice contributions to total aerosol optical depth, and showed  
475 systematic variations in water-ice cloud abundance throughout day and night (Smith et al.,  
476 2025b). TIRS  $\tau_{\text{ice}}$  retrievals demonstrated substantial consistency with the diurnal patterns  
477 observed by EMIRS over Syrtis Major near Perseverance (e.g, Figure 7 from Smith et al., 2025).  
478 During aphelion season, TIRS observed a strong evening peak in  $\tau_{\text{ice}}$ , with values above 0.4 and  
479 often exceeding 0.5, and higher nighttime water-ice abundance compared to daytime values.  
480 Average EMIRS  $\tau_{\text{ice}}$  in this region during aphelion season was in the range of 0.4 to 0.7 in the  
481 late evening between LTST 18 and 24. Substantial water-ice cloud formation was also observed  
482 during less dusty periods outside of aphelion season, with a pronounced morning peak that often  
483 occurred shortly after dawn and a separate late evening peak.

484 The Emirates eXploration Imager (EXI) onboard EMM provides complementary observations of  
485 water-ice clouds to EMIRS through UV and visible wavelength images of the Martian disk  
486 during daytime hours (Jones et al., 2021; Wolff et al., 2022). While EXI retrievals of  $\tau_{\text{ice}}$  at 320  
487 nm are not directly comparable in magnitude to EMIRS thermal infrared retrievals without  
488 consideration of additional water-ice particle properties such as effective radius, relative changes  
489 across spatial and temporal scales can be compared between instruments. EXI  $\tau_{\text{ice}}$  retrievals also  
490 cannot distinguish between surface and atmospheric ice, and a surface ice mask was not applied  
491 to these results, leading to larger optical depths over the polar ice caps. Nevertheless, the two  
492 instruments showed general agreement in their observations of water-ice cloud variations across  
493 the diurnal cycle, particularly for the low- and mid-latitude clouds this analysis is focused on.

494 Figure 9 shows EMIRS and EXI  $\tau_{\text{ice}}$  retrievals for five periods spanning different seasonal  
495 conditions during MY 36 and 37. During aphelion seasons, both instruments captured key  
496 features of the ACB, including the pronounced morning peak in cloud abundance and subsequent  
497 midday minimum. The latitudinal extension of early morning ACB clouds beyond typical  
498 daytime bounds was clearly evident in both datasets. Notably, during aphelion season both  
499 instruments showed the morning peak occurring shortly after sunrise, consistent with the  $\sim 75^\circ$   
500 solar incidence angle relationship discussed previously. During less dusty periods, similar  
501 patterns of diurnal cloud development were observed, though with reduced magnitude. The polar  
502 regions showed higher  $\tau_{\text{ice}}$  in EXI observations compared to EMIRS—which may be partially  
503 attributed to surface ice effects or smaller particle sizes, though increased coverage further  
504 poleward in EXI results and lower thermal contrast and radiance in EMIRS observations over the  
505 poles may explain some of the discrepancies as well.



**Figure 9** Zonal mean water-ice optical depth across the diurnal cycle for EMIRS (rows 1 and 3) and EXI (rows 2 and 4) retrievals across five periods of interest during MY 36 & 37. Red contour lines in EMIRS results show the  $90^\circ$  mean solar incidence angle (the day-night terminator). Note that EXI and EMIRS optical depths are referenced to different wavelengths and are not directly comparable in magnitude.

506 The consistency between EMIRS observations and both surface-based and orbital measurements  
 507 provides additional validation of the diurnal behavior of water-ice clouds across different spatial  
 508 scales and observational techniques. All three instruments captured the fundamental pattern of  
 509 morning and evening peaks in cloud abundance separated by midday minima and higher  
 510 nighttime  $\tau_{ice}$  values, as well as persistent cloud activity during less dusty periods outside of  
 511 aphelion. The timing of the morning peak shortly after dawn was particularly well-documented

512 across instruments, occurring near  $75^\circ$  solar incidence angle in EMIRS results, and occurring  
513 past sunrise in EXI and TIRS observations.

## 514 **6 Conclusions**

515 Understanding the full diurnal cycle of Mars' water ice clouds, particularly during nighttime  
516 hours when previous observations have been limited, is crucial for characterizing their role in the  
517 Martian climate system. This study presents the first detailed observations of the full diurnal  
518 cycle of water-ice clouds in the Martian atmosphere using thermal infrared spectra from EMIRS.  
519 The unique orbital configuration of EMM allowed for comprehensive observational coverage  
520 across all local times and nearly all latitudes and longitudes for nearly two Martian years.  
521 Updates to the retrieval algorithm, including improved treatment of surface thermophysical  
522 effects and inclusion of the  $45\text{-}\mu\text{m}$  water ice band available in EMIRS observations, enabled  
523 retrieval of water ice optical depth throughout the full diurnal cycle. The result is a new  $\tau_{\text{ice}}$   
524 observational dataset with broad spatial and temporal coverage that includes nighttime hours.

525 Key findings of this study include:

- 526 • A persistent diurnal pattern in low- and mid-latitude clouds characterized by a midday  
527 minimum in  $\tau_{\text{ice}}$ , with relative increases during both evening and early morning hours.  
528 The morning peak, which was often the highest optical depth in the diurnal cycle,  
529 consistently occurred shortly after dawn near a  $75^\circ$  solar incidence angle, across a range  
530 of latitudes and seasons.
- 531 • Across an aphelion season average, zonal mean  $\tau_{\text{ice}}$  ranged from a maximum of  $\sim 0.5$  in  
532 the morning to a minimum of  $\sim 0.15$  near midday—consistent with previous TES midday  
533 observations around 14:00 LTST. Higher water-ice optical depths were observed at night  
534 compared to midday throughout much of the year, with particularly pronounced  
535 differences during aphelion season when ACB clouds were often more than twice their  
536 midday minimum values.
- 537 • Substantial spatial variability was observed in nighttime cloud patterns, with evening  
538 clouds at low- and mid-latitudes distributed across more longitudinal regions, while the  
539 most abundant early morning clouds tended to occur over the Tharsis region. The ACB  
540 reached considerably wider latitudinal extents in the early morning hours, particularly  
541 over Tharsis where clouds extended from roughly  $25^\circ\text{S}$  to  $45^\circ\text{N}$ —well beyond typical  
542 midday bounds.
- 543 • Detection of considerable water ice cloud abundance outside of aphelion season during  
544 less dusty periods between regional dust storms—particularly at night and during  
545 morning and evening peaks. Mean  $\tau_{\text{ice}}$  values of  $\sim 0.2$  were commonly observed during  
546 these periods, revealing cloud activity that was often not apparent in previous datasets  
547 limited to near-midday observations.

548 Comparisons with TIRS and EXI observations showed good agreement in the diurnal patterns of  
549 cloud development, providing similar observations of key features including the post-dawn  
550 morning peak in cloud abundance. The observed diurnal patterns also showed notable  
551 similarities to those expected from the influence of atmospheric thermal tides on cloud formation  
552 conditions, though detailed investigation of this relationship requires further study. The  
553 comprehensive diurnal coverage presented here offers new constraints on water-ice cloud

554 abundance for atmospheric models and an improved understanding of the role of water-ice  
 555 clouds in the Martian climate system. Future work will focus on detailed investigation of the  
 556 drivers of cloud development, including how thermal tides and topography modulate cloud  
 557 formation and create feedbacks with atmospheric temperatures, and the implications for  
 558 atmospheric transport and the global water cycle.

## 559 Acknowledgments

560 The material is based upon work supported by NASA under award number 80GSFC24M0006.  
 561 We would also like to acknowledge and thank the reviewers for their helpful suggestions  
 562 regarding descriptions of EMIRS uncertainties and classification of the periods of interest.

## 563 Open Research

564 Data from the Emirates Mars Mission and EMIRS are freely and publicly available on the EMM  
 565 Science Data Center (SDC, <http://sdc.emiratesmarsmission.ae>). Relevant results from this work,  
 566 including retrieval output and uncertainty weighted mean values for water-ice optical depths,  
 567 have been saved as NetCDF4 files and archived on Mendeley Data (Atwood, 2025).

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