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Key Points:

- We propose a recalibration of InSight pressure data to correct an unexpected sensitivity to the sensor temperature
- A comparison between the InSight and Viking 1 pressure data does not show variations larger than ± 8 Pa in the global atmospheric mass
- This comparison also supports the absence of long-term variability in the dynamics of seasonal cap formation and sublimation

Supporting Information:

Supporting Information may be found in the online version of this article.

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InSight Pressure Data Recalibration, and Its Application to the Study of Long-Term Pressure Changes on Mars

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Abstract Observations of the South Polar Residual Cap suggest a possible erosion of the cap, leading to an increase of the global mass of the atmosphere. We test this assumption by making the first comparison between Viking 1 and InSight surface pressure data, which were recorded 40 years apart. Such a comparison also allows us to determine changes in the dynamics of the seasonal ice caps between these two periods. To do so, we first had to recalibrate the InSight pressure data because of their unexpected sensitivity to the sensor temperature. Then, we had to design a procedure to compare distant pressure measurements. We propose two surface pressure interpolation methods at the local and global scale to do the comparison. The comparison of Viking and InSight seasonal surface pressure variations does not show changes larger than ± 8 Pa in the CO₂ cycle. Such conclusions are supported by an analysis of Mars Science Laboratory (MSL) pressure data. Further comparisons with images of the south seasonal cap taken by the Viking 2 orbiter and MARCI camera do not display significant changes in the dynamics of this cap over a 40 year period. Only a possible larger extension of the North Cap after the global storm of MY 34 is observed, but the physical mechanisms behind this anomaly are not well determined. Finally, the first comparison of MSL and InSight pressure data suggests a pressure deficit at Gale crater during southern summer, possibly resulting from a large presence of dust suspended within the crater.

Plain Language Summary Observations of the permanent CO₂ ice cap at the south pole of Mars in the 2000s suggested that the cap was eroding, possibly releasing a significant amount of CO₂ into the atmosphere. To test this hypothesis, we compare surface pressures recorded by Viking in the 1970s and those recorded by InSight in 2018–2021 to confirm or refute the suspected increase of the atmospheric mass. After establishing our comparison method, we correct the influence of the sensor temperature on the InSight pressure data, which was discovered during our investigation. Comparison of the pressure data, as well as images of the seasonal caps taken by orbiters, do not reveal any change in the atmospheric mass or the dynamics of the seasonal ice caps that develop during the martian year. These conclusions are reinforced by reanalyzing the pressure data recorded by the Curiosity rover. Only small interannual changes are observed, potentially related to the effect of the dust storms that happened on Mars between 2016 and 2018. Finally, we report a possible pressure deficit at MSL's location during southern hemisphere summer, potentially explained by the unusual presence of dust in the crater air.

1. Introduction

The retreat of the Southern Seasonal Polar Cap (SSPC) during local summer leaves a residual perennial deposit mainly composed of CO₂ ice (Kieffer et al., 1972). This deposit, known as the South Polar Residual Cap (SPRC) can persist during the Southern summer because of its high albedo (Jakosky & Haberle, 1990) and the exchanges with the CO₂ ice at the surface and the CO₂ ice deposits that are buried through permeable ground water ice at this location (Buhler et al., 2020). The SPRC is one of the CO₂ reservoirs that can significantly affect the atmospheric mass through sublimation or deposition (Leighton & Murray, 1966). The stability of this reservoir over time is a long-standing debate in martian climate science. While Blackburn et al. (2010) predicted the disappearance of

the SPRC within a few years, Piqueux and Christensen (2008) reported limited changes in the extent and ice-covered area of the SPRC since the Mariner 9 mission in 1972 and telescopic observations in the twentieth century. However, Piqueux and Christensen (2008) were unable to retrieve a mass balance of the cap. Other monitoring of the SPRC surface since the Mariner 9 and Viking era led to mass balance estimates suggesting either an erosion of the SPRC (Blackburn et al., 2010; Malin, Bell, et al., 2001; Thomas et al., 2009, 2013, 2016) or a possible ice accretion (Thomas et al., 2016). This presumed erosion or accretion of the SPRC open the possibility of secular pressure changes on Mars: if the SPRC loses CO₂ ice year after year, the sublimated CO₂ ice goes into the atmosphere, increasing its global mass and thus also the global mean surface pressure on Mars (Haberle & Kahre, 2010; Kahre & Haberle, 2010; Malin, Bell, et al., 2001; Thomas et al., 2016). Conversely, the deposition of CO₂ ice on the cap would decrease the atmospheric mass, and thus the atmospheric pressure. Observations from Malin, Bell, et al. (2001) suggested that the CO₂ ice thickness decreases by nearly 0.4 m per Martian Year (MY). After computing the total volume of ice that is eroding, and assuming a CO₂ ice density of $1.6 \times 10^3 \text{ kg m}^{-3}$, they estimated that the amount of CO₂ released into the atmosphere, and thus the increase of surface pressure, is nearly +13 Pa per Martian Decade (MD). Blackburn et al. (2010) also estimated a possible increase of surface pressure between +0.5 and +13 Pa/MD. A recent study by Thomas et al. (2016) qualified the two previous estimations by reporting a much smaller variation of SPRC mass balance, with a possible variation of surface pressure between −2.3 Pa/MD and +1.6 Pa/MD.

In addition to this possible secular change in atmospheric mass, we can investigate the possibility of an unknown mechanism that would change the dynamics of formation and sublimation of the CO₂ seasonal caps. Thermal infrared observations (e.g. (Piqueux et al., 2015)), spectroscopic studies (e.g. (Brown et al., 2010; Langevin et al., 2007)), or cap albedo monitoring (e.g., (Calvin et al., 2015; Calvin et al., 2017)) have already reported interannual variability in the formation and recessions of the seasonal caps. However, no comparison has been made with the present observations (i.e., for martian year 35–36).

A direct way to assess long-term pressure changes on Mars consists of comparing surface pressure measurements separated by several martian decades. By this method, we can check if the atmospheric mass has changed, and study possible variability in the dynamic of the seasonal ice caps. Such a comparison must be done carefully, however, because of the influence of orography and meteorological variability on the annual surface pressure cycle (Hourdin et al., 1993, 1995). The comparison of pressure measurements made by Viking between 1976 and 1982 and those by Phoenix in 1997, after being corrected for both topography differences and atmospheric dynamics simulated by a Global Circulation Model (GCM), showed a possible 10 Pa rise of surface pressure (Haberle & Kahre, 2010) which corresponds to 5 Pa/MD. However, the combined uncertainties in both the measurements and the interpolation methodology were not sufficiently accurate to draw any conclusions about a secular pressure change. More recently, the comparison between Mars Science Laboratory pressure measurements, which have been recorded since 2012, and Viking measurements did not show significant changes in surface pressure (Haberle et al., 2014). However, these conclusions are limited by the sensitivity to the hydrostatic adjustment of surface pressure as the rover is climbing Mount Sharp in Gale Crater (Haberle et al., 2014; Richardson & Newman, 2018), and the sensitivity of the atmospheric dynamics at Gale Crater that have to be resolved by a mesoscale model (Pla-Garcia et al., 2016; Rafkin et al., 2016). Hence, the analysis of the first available surface pressure data neither confirmed nor denied any long-term pressure changes. Remote sensing measurements of surface pressure could be interesting to exploit (e.g. (Forget et al., 2007; Withers, 2012)) but these measurements are not accurate enough compared to in-situ surface pressure measurements and will not be exploited here.

In 2018, the InSight mission deployed a geophysical and meteorological observatory, including a pressure sensor, at the surface of Mars (Banerdt et al., 2020; Banfield et al., 2019). The instruments are deployed on a static lander at Elysium Planitia, a relatively flat area located at 4.5°N, 135.6°E (Golombek et al., 2020), thus reducing the sensitivity of surface pressure measurements to both hydrostatic adjustment and atmospheric dynamics. Pre-flight calibration and tests suggested that the performances of the sensor were good enough to detect changes in the atmospheric mass and CO₂ cycle (Spiga et al., 2018). We thus propose in this paper to compare the InSight pressure data with the Viking pressure data to assess the possibility of long-term pressure changes over two martian decades.

We present in Section 2 the methodology of the pressure interpolation that will lead our comparison between Viking and InSight data. A closer look at the InSight data reveals a calibration problem due to a sensor temperature sensitivity. We propose an empirical recalibration and test the reliability of this correction in Section 3.

The comparison between the InSight corrected pressure data and Viking surface pressure measurements is then presented in Section 4 to check for a possible secular increase or decrease of atmospheric mass. Long-term variations in the dynamics of the seasonal ice caps between the 1970s and 2018 are also investigated using pressure data and satellite images from the Viking and InSight eras, respectively. We then extend the scope of this study by also exploiting the Phoenix and MSL measurements to detect any evolution of the atmospheric mass in Section 5. We also look at the possible influence of interannual variability of the seasonal cap due to the dust cycle. The main conclusions from our investigation are presented in Section 6.

2. Methodology of Pressure Interpolation

The interpolation of the Viking pressure to the InSight landing site requires taking into account planetary-scale atmospheric dynamics that affect the surface pressure (Hourdin et al., 1993, 1995). Even local interpolation between two close points must include the influence of local weather phenomena, like slope winds. Hence, interpolating pressure cannot be limited to integrating the hydrostatic equation to correct for altitude differences. To take into account the impact of atmospheric dynamics at all scales, we propose two high-accuracy interpolation methods: one on a local scale (typically within a crater, a slope, etc.), and one on a regional-to-global scale.

2.1. Local Pressure Interpolation

We consider here a local situation in which two points are close enough so that large-scale dynamic pressure gradients related to the global atmospheric circulation and regional flows can be neglected. Let us consider two points A and B located at different altitudes (Figure 1a). Since these two points are close, the main factor that impacts the difference in the absolute pressure is altitude, thus we could assume hydrostatic equilibrium and recast pressure (P_B) at point B to the altitude at point A ($P_{B \rightarrow A}$) with:

$$P_{B \rightarrow A} = P_B e^{-\frac{z_A - z_B}{H}} \quad (1)$$

where z corresponds to the altitude of each point, H is the scale height expressed as:

$$H = \frac{RT}{\mu g} \quad (2)$$

with $R = 8.3145 \text{ J kg}^{-1} \text{ mol}^{-1}$ the molar gas constant, T is the mean atmospheric temperature between A and B weighted by vertical pressure field (in Kelvin), $\mu = 43.34 \times 10^{-3} \text{ kg mol}^{-1}$ the mean molecular weight of Mars atmosphere and $g = 3.72 \text{ m s}^{-2}$ Mars' surface gravity.

On terrains with an uneven topography, local circulations, like slope winds, can appear as a consequence of temperature gradients. Hence, the choice of a scale height H , and thus the temperature to take into account in Equation 2, is important to consider the major effect of these local circulations (Forget et al., 2007; Spiga et al., 2007): the temperature choice in H will indicate which path should be used to integrate the hydrostatic equation (red and green lines in Figure 1a). Such effects are very important for the Mars Science Laboratory mission for instance. As the Curiosity rover moves in Gale Crater, with significant gains of elevation (several hundred meters), local circulations and slope winds (Pla-Garcia et al., 2016; Rafkin et al., 2016; Richardson & Newman, 2018) also contribute to the absolute pressure recorded by the rover. Forget et al. (2007); Spiga et al. (2007) suggested using the temperature at an altitude of 1 km above the surface in Equation 2 to take into account the effect of slope winds at the GCM scale, while Ordonez-Etxeberria et al. (2019) used the air temperature at an altitude of 2 m when using MSL pressure data. Haberle et al. (2014) also questioned the choice of the scale height H that has to be used when exploiting MSL data. Their study of the sensitivity of pressure data to this scale height shows that, with extreme temperature scenarios, the absolute pressure can be influenced by several Pascals. However, they never determine which scale height is the right one to use with these data.

Thus, we investigate here the scale height that better matches MSL observations, and quantify the errors made during the interpolation of the surface pressure, using the example of Gale Crater. To do so, simulations of Gale Crater with the LMD mesoscale model (Spiga & Forget, 2009) were performed. The domain for the simulations ranges from 22°S to 30°N and 108°E to 163°E, with a spatial resolution of 0.16°, including thus the InSight landing site, Gale Crater and its circulation.

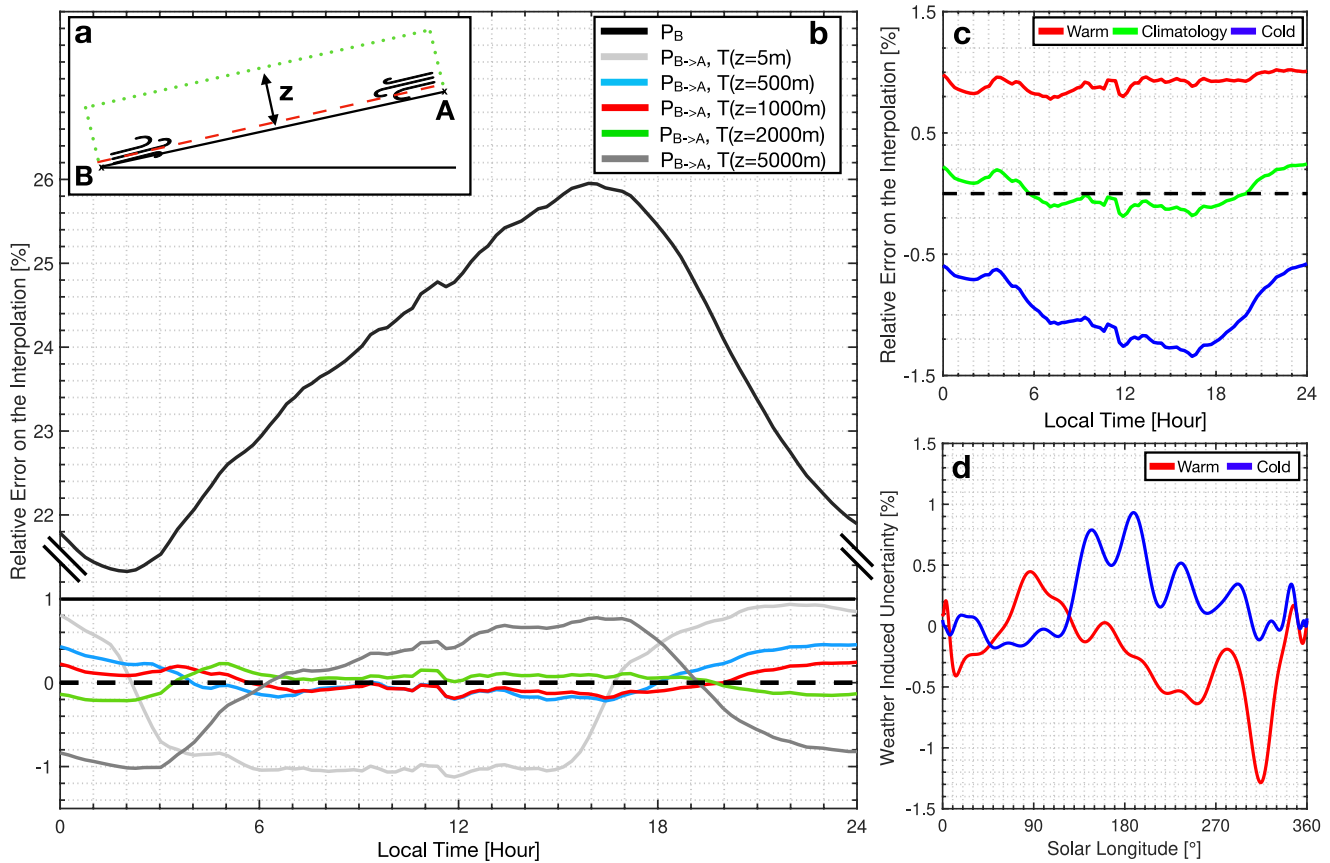


Figure 1. (a) Schematics of the problem of interpolation with slope winds between the bottom of Gale Crater (point B) and the rim of the crater (point A). Colored dots illustrate the different paths that can be taken to integrate the hydrostatic equation. (b) Relative error of the interpolated pressure from point B to point A and the exact pressure at (a). The black curve is the relative error when point B is not interpolated to point A, while colored curves are for the relative errors when using different altitudes for the temperature. The air temperatures T used in the interpolations are computed at an altitude z above point (b) (c) Relative error on the local interpolation when using the temperature at 1 km above the surface at point A when considering several kinds of weather scenarios. Results are given for the mesoscale simulation that ran for $L_s = 180^\circ$, but similar results (i.e., same magnitudes) are obtained at other L_s . (d) Weather-induced uncertainty of the Viking surface pressure interpolated to InSight landing site computed with extreme dust scenarios when compared to *clim* dust scenario (red and blue curves).

We take the diurnal cycle of surface pressure simulated at several seasons at grid points at the bottom (B) and the rim of Gale Crater (A) ($\Delta z = 1725\text{m}$) in the direction of MSL trajectory). We then interpolate the pressure at point B (P_B) to location A using several altitudes for the temperature above the point B to compute H . We then compute the relative error between the exact modeled pressure at A (P_A), and the interpolated pressure from B to A ($P_{B \rightarrow A}$). Results are presented in Figure 1b. They show that choosing the temperature at an altitude between 500 m and 2 km above the surface is better to take into account the effect of local dynamics on the pressure interpolation as it minimizes the relative difference between P_A and $P_{B \rightarrow A}$. In the following, we choose to compute the scale height H by using the temperature at an altitude of 1 km. When interpolating actual measurements, this temperature is not available from observations and instead has to be estimated using an atmospheric model. The main uncertainty in this calculation results from the sensitivity of temperatures to the dust opacity, which is not perfectly known. To check the sensitivity of the interpolation to these weather conditions and thus to an error in T , we use the GCM runs to quantify the impact of the dust opacity using three dust scenarios as input:

1. A climatology (*clim*) scenario, derived by averaging the available observations of dust from MY 24, 25, 26, 28, 29, 30, and 31 outside the global dust storm period (Montabone et al., 2015). This scenario represents a nominal dust scenario in the absence of major dust storms.
2. A cold scenario which corresponds to an extremely clear atmosphere. At a given date and location, the dust opacity is set to be the minimum observed over MY 24–31, and further decreased by 50%.

3. A *warm* scenario which corresponds to “dusty atmosphere” conditions, outside of global dust storms. The dust opacity at a given location and date is set to the maximum observed over MY 24–31, excluding the periods of the MY 25 and MY 28 global dust storms, and further increased by 50%.

These scenarios are used in the Mars Climate Database (MCD, Millour et al. (2018)), and frame well the different temperature observations made by several spacecrafts at a $3 - \sigma$ level (Millour et al., 2018) ($3 - \sigma$ means that 99.7% of the excursions of the value from the mean are under three standard deviations). Using these scenarios, our simulations show that the temperature of the air at an altitude of 1 km can vary by several kelvins. We consider the worst-case scenario, assuming that the *cold* scenario decreases the temperature by 10 K compared to the *clim* scenario; and the *warm* scenario increases the temperature by 10 K (simulations report a maximum of ± 8 K in terms of anomaly, and we add a 2 K margin). The relative errors made in the interpolation by using these temperature deviations instead of nominal temperatures are presented in Figure 1c. This sensitivity study shows that the relative $3 - \sigma$ accuracy of this interpolation method is almost 1%, and is thus acceptable to exploit MSL pressure data. In summary, we found that an accurate way to interpolate surface pressure from a point B to a point A at a local scale consists of using Equation 1 with the scale height computed using the temperature at an altitude of 1 km above point B (Equation 2).

2.2. Large-Scale Pressure Interpolation

At the planetary scale, in addition to the hydrostatic adjustment and local dynamic effects, we must take into account large-scale dynamic pressure gradients in the interpolation (Hourdin et al., 1993, 1995). Hence, interpolating Viking pressure to InSight cannot be done by using Equation 1 alone, as it does not consider these gradients.

To account for these atmospheric large-scale dynamic components, we use a method based on the LMD GCM (Forget et al., 1999; Hourdin et al., 1993). Practically, the interpolation of Viking pressure data to obtain the pressure to any location on Mars consists of determining the spatial variation of surface pressure in the GCM, with interpolation from the coarse GCM topography grid (5.625° in longitude, 3.75° in latitude) to the high-resolution MOLA grid (32 pixels per degree), plus a correction to perfectly match the seasonal variations at the Viking 1 site.

Thus, the interpolation of Viking 1 surface pressure at any location, P_s , is done with (Forget et al., 2007; Spiga et al., 2007):

$$P_s = \langle P_{\text{Viking}} \rangle \frac{P_{\text{GCM}}}{\langle P_{\text{GCM, Viking}} \rangle} e^{-\frac{z - z_{\text{GCM}}}{H}} \quad (3)$$

where P_{GCM} is the pressure predicted by the GCM at the site we want to interpolate to, $\langle P_{\text{Viking}} \rangle$ corresponds to the pressure records of Viking averaged over 15 sols to remove any weather variations (thermal tides and transient waves), $\langle P_{\text{GCM, Viking}} \rangle$ is the surface pressure predicted by the GCM at the location of Viking 1 and also smoothed over 15 sols. $z - z_{\text{GCM}}$ is the difference between the MOLA altitude and the altitude defined with the interpolation of the GCM topography grid at the location we consider, and H corresponds to the scale height computed with Equation 2 using the air temperature at the site we want to interpolate the Viking measurements. The same procedure as the one used in Section 2.1, using the GCM pressure field binned every 2 hr, again shows that we must consider the temperature at an altitude of 1 km above the surface. In this expression (Equation 3), $\langle P_{\text{Viking}} \rangle$ is the pressure we want to interpolate, $\frac{P_{\text{GCM}}}{\langle P_{\text{GCM, Viking}} \rangle}$ is the correction of atmospheric dynamics induced by the pressure gradients, and $e^{-\frac{z - z_{\text{GCM}}}{H}}$ is a hydrostatic correction, taking into account the effect of local dynamics.

We use in this study the Viking 1 surface pressure data rather than Viking 2 data. Viking 1 data are indeed more complete after removing the measurements made during dust storms, less sensitive to baroclinic activity (Ryan & Sharman, 1981; Tillman, 1988; Tillman et al., 1993), and closer to InSight than Viking 2 (Golombek et al., 2020; Morris & Jones, 1980), thus limiting the sensitivity to errors in the correction of the dynamics of the atmosphere.

The uncertainty of the interpolation depends on two independent uncertainties: one linked to the Viking 1 pressure measurements and one to the weather-induced uncertainty. On the one hand, pre-flight tests showed that the precision of the Viking pressure sensors was better than $\pm 0.2\%$ of the readings, plus a term due to a temperature

dependency of nearly 0.18% (Mitchell, 1977; Seiff & Kirk, 1977). Consequently, the precision of the pressure measurements was ± 3.4 Pa for Viking 1. Such errors in the precision can be mitigated, however, as we are using a pressure signal averaged over 15 sols. Assuming that this precision error on a single measurement can be modeled by white noise with a $3 - \sigma$ of 3.4 Pa, we can reduce the uncertainty on the diurnal average pressure value by a factor \sqrt{N} where N is the number of measurements used for the diurnal or 15 sols average. Typically, 200 measurements per sol are used to compute the diurnal average (Barnes, 1980). Therefore, by using a 15 sols averaged surface pressure in Equation 3, the relative sensitivity to the uncertainty of Viking measurements is limited to 0.06 Pa, and is thus completely negligible in the following.

On the other hand, Viking measurements are also impacted by systematic errors due to the instrumental drift, the 8-bit telemetry resolution, and the uncertainty on the zero level of the pressure sensor's output voltage. Based on the apparent stability of the sensor because of the great repeatability of the pressure data from one year to another outside dust storm periods (Hess et al., 1980; Tillman et al., 1993), the instrumental drift had been estimated to be only -0.1 ± 1 Pa per Earth year and will be neglected in the following. One could assume that long-term atmospheric changes can be compensated by the sensor drift, canceling thus the interannual variability in the Viking measurements. This assumption has been ruled out as it appears very unlikely, based on the accurate agreement with the mean surface pressure and harmonic analysis (Hess et al., 1980; Tillman et al., 1993). The error due to the 8-bit telemetry resolution (Seiff, 1976; Tillman et al., 1993) yields an uncertainty of at most 8.8 Pa in the absolute pressure level for one single measurement, even if the sensor has a theoretical resolution of nearly 1 Pa (Hess et al., 1976; Seiff & Kirk, 1977). Assuming that this uncertainty on a single measurement can be modeled by white noise with a $3 - \sigma$ of 8.8 Pa, and using a 15 sols averaged surface pressure, this uncertainty is reduced to 0.16 Pa and will also be neglected in the following. The last systematic error related to Viking data is due to the uncertainty on the zero level of the pressure sensor's output voltage. This was determined by readings taken just before atmospheric entry. The resolution uncertainty in these zero readings causes an uncertainty of up to 8.8 Pa in the absolute pressure level (Kahanpää et al., 2021; Seiff & Kirk, 1977). Hess et al. (1980) proposed adding 4.4 Pa to each measurement as it should be the best estimate of the true pressure measurements, reducing the absolute error by half. However, it remains unclear if this adjustment has been implemented in the Planetary Data System (PDS) where data are available. We will thus consider in the rest of the study that the absolute error on Viking 1 pressure measurement is $\Delta P_{\text{Viking}} = 8.8$ Pa at a $3 - \sigma$.

The second uncertainty in the interpolation is the influence of weather conditions, which impacts T and thus H in Equation 3 as well as the pressure predicted in the GCM. To study the impact of these conditions on the GCM output, we compute the interpolation of Viking 1 pressure to the InSight landing site by using the three dust scenarios described above, as they bracket well the possible states of the atmosphere (Millour et al., 2018). We then compute the weather-induced uncertainty, defined as the relative difference between the pressure at InSight's location derived with the extreme dust scenarios (*cold* and *warm*) and the *clim* dust scenario (Figure 1d). Figure 1d underlines that this weather-induced uncertainty is generally limited to 1% of the absolute pressure at $3 - \sigma$. We set this uncertainty as 1% of the mean annual pressure of 700 Pa at InSight's landing site (Figure 2), that is, by $\Delta P_{\text{weather}} = 7$ Pa at $3 - \sigma$. It should be noted that we use dust scenarios derived from Mars Climate Sounder (MCS, McCleese et al. (2007)) observations from MY 29 to MY 35 (Montabone et al., 2015, 2020) for our comparisons. The weather-induced uncertainty is therefore much lower as these accurate observational scenarios help to compute T , and thus H , in Equation 3 precisely.

Combining the independent uncertainty of Viking measurements ΔP_{Viking} and the weather-induced uncertainty $\Delta P_{\text{weather}}$ yields an uncertainty of the interpolation of nearly $\sqrt{8.8^2 + 7^2} \approx 11.2$ Pa at $3 - \sigma$. Such a threshold is at the limit of the lowest predictions of atmospheric mass variations possibly indicated by cap studies (Thomas et al., 2016), but well below the first estimates made at the beginning of the 2000s (Blackburn et al., 2010; Malin, Bell, et al., 2001).

3. Recalibration of InSight Pressure Data

The InSight pressure sensor is located on the lander deck at a height of approximately 1.2 m, with a sampling rate of up to 20 Hz and a noise level lower than 10 mPa, which theoretically represents an unprecedented quality compared to the different pressure sensors that have operated on the surface of Mars (Banfield et al., 2019, 2020; Spiga et al., 2018). These data are calibrated by using output voltage and pressure sensor temperature channels.

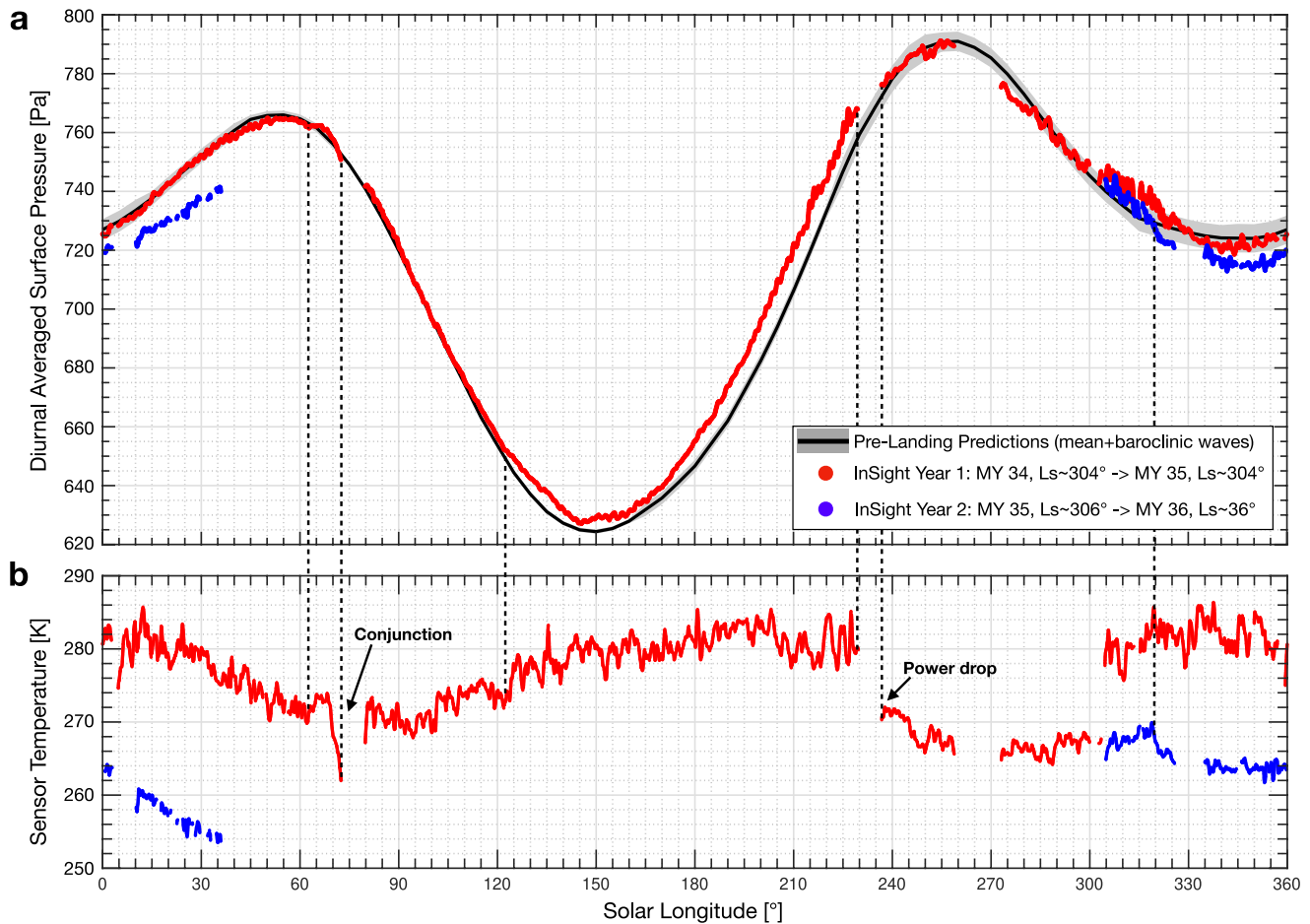


Figure 2. (a) Diurnal averaged surface pressure computed from the 20 Hz data acquired during the two years of the mission (red and blue), with pre-landing surface pressure predictions (black curve) and baroclinic waves amplitudes (gray filled area) from Spiga et al. (2018). (b) Diurnal averaged temperature of the pressure sensor. Red dots are for the first year of the mission, while blue dots represent the measurements taken during the second year of the mission. Dashed black lines highlight the significant correlations between the pressure sensor temperature and the pressure measurements.

We use in this study the 20 Hz data and average them with a 50s window to remove any effects of high-frequency pressure events (e.g., Chatain et al. (2021); Spiga et al. (2021)). We then compute the diurnal average of these signals. To do so, we interpolate the data from previous and following sols to complete the diurnal cycle when there are small gaps (typically of a few hours) in the data. We then interpolate the measurements onto a regular temporal grid containing 100 points per sol. From these interpolated points, we compute the diurnal average. The diurnally averaged surface pressure obtained for the entire data set (~1.25 MY) is presented in Figure 2a.

3.1. Sensor Temperature Sensitivity of the InSight Pressure Data

A direct comparison between the pressure measurements made one year apart during the first and second martian years of the InSight mission shows a large difference (Figure 2). This cannot be explained by the instrumental drift reported in Banfield et al. (2019); Spiga et al. (2018) or by any likely major meteorological event as no significant long-lived global events have been observed. This difference is also not observed by MSL pressure measurements (Figure S1), thus questioning the reliability of the InSight pressure measurements. Furthermore, the divergence between the measurements made two years apart seemed to increase toward the end of the mission, when the power allocated to the pressure sensor was very low because of the accumulation of dust on the solar panels, leading to a decrease of the pressure sensor temperature. In the following, sensor temperature will refer to the temperature of the pressure sensor and does not refer to other temperatures (e.g., TWINS air temperature (Banfield et al., 2019) or other sensor temperatures measured by InSight).

A close comparison of the pressure measurements and sensor temperature (Figure 2b) reveals that the pressure measurements are very likely to be affected by some drops or rises in the sensor temperature. An illustration of this correlation is at $L_s \sim 72^\circ$, just before conjunction. Sensors were powered off, after the warm Electronics Box had cooled off, and a nonphysical drop occurred in the pressure measurements. This correlation questions the reliability of the calibration of the absolute pressure data to the sensor temperature. The pre-flight calibration of the pressure sensor with temperature may not be as accurate as expected, possibly because of the existence of temperature gradients within the instrument under real martian conditions. Such an effect had already been identified as responsible for pressure measurement errors on Phoenix (Taylor et al., 2010). It is important to note that most of the scientific results obtained from the pressure data are not impacted by this calibration problem. These works (see for instance Banerdt et al. (2020); Spiga et al. (2021); Chatain et al. (2021)) use relative pressure variations and not absolute measurements, and at high frequencies. At these frequencies, that is, for timescales of the order of a sol, or less, the fluctuations of the sensor temperature are negligible. The calibration problem detected is thus nullified when using relative variations of measured pressures, and therefore does not bias the scientific results obtained. We propose in Section 3.3 to correct this thermal effect using MSL pressure data.

3.2. Mars Science Laboratory Pressure Data

In the absence of major meteorological events, we can expect limited interannual variations between the pressure measured during the first and the second year of the InSight mission (see legend of Figure 2 for definitions). de la Torre Juárez et al. (2019) reported a strong interannual variability of the pressure data at the end of MY 34 and the beginning of MY 35 compared to other years using MSL pressure data (Figure S1). Such a difference might be linked to the global dust storm of MY 34 that possibly had an impact on the extent of the NSPC.

To take into account a possible interannual variability between InSight's first and second years of the mission, we use MSL Rover Environmental Monitoring Station (REMS) pressure data available in the PDS. The calibrated data extracted from the PDS extend from MY 31, at a solar longitude (L_s , the Mars-Sun angle, measured from the Northern Hemisphere spring equinox where $L_s = 0^\circ$) $L_s \sim 150^\circ$, to MY 36, $L_s \sim 21^\circ$. The REMS pressure sensor acquires data during the first five minutes of each Local Mean Solar Time (LMST) with additional hour-long acquisitions that cover a full diurnal cycle about every 6 sols (Gómez-Elvira et al., 2014). To take into account the vertical displacement of the rover on pressure measurements that could account for more than 15 Pa in the pressure records (Ordóñez-Etxeberria et al., 2019), we interpolate the pressure data from the position of the rover, determined with the Ancillary Data Record (ADR) files, to the MSL landing site using the method described in Section 2.1, Equation 1. The air temperatures at an altitude of 1 km above the surface are computed with the MCD, using dust scenarios from Montabone et al. (2015); Montabone et al. (2020) as inputs for the simulations. The results of this interpolation are presented in Figure S1. The correction of these altitude differences can reach 20 Pa (more than Ordóñez-Etxeberria et al. (2019) because we use a larger data set). We obtain a data set with little interannual variability, except during the end of MY 34 and the beginning of MY 35, as noted by de la Torre Juárez et al. (2019). This good repeatability in our corrected MSL pressure data set further validates our interpolation method presented in Section 2.1 to correct altitude differences.

3.3. Recalibration of the Pressure Measurements

We define $E(T(t))$ (in Pa) as the sensitivity of the pressure measurements with regards to the sensor temperature T . The corrected measured pressure $P_{\text{InSight, Corrected}}$ at a time t can be written as:

$$P_{\text{InSight, Corrected}}(t) = P_{\text{InSight, Measured}}(t) + E(T(t)) \quad (4)$$

We average the pressure measured by MSL and InSight over 15 sols to eliminate the contribution of any dynamical component like thermal tides and baroclinic activity. These averaged pressure values are denoted $\langle P \rangle$ in the following. As InSight and MSL are relatively close (~ 600 km), we assume that the correction of the large-scale atmospheric dynamics between the two sites can be neglected. Our simulations show indeed that this correction is limited to 1 Pa at $3 - \sigma$ and is thus negligible.

During InSight year 1 (Y_1 : MY 34, $L_s \sim 304^\circ$ to MY 35, $L_s \sim 304^\circ$) and year 2 (Y_2 : MY 35, $L_s \sim 306^\circ$ to MY 36, $L_s \sim 36^\circ$) of the mission, we have:

$$\begin{cases} \langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle = \langle P_{\text{MSL}}(t_{Y_1}) \rangle e^{-\frac{\Delta z}{H(t_{Y_1})}} \\ \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle = \langle P_{\text{MSL}}(t_{Y_2}) \rangle e^{-\frac{\Delta z}{H(t_{Y_2})}} \end{cases} \quad (5)$$

with Δz the difference of altitude between the InSight and MSL landing site, and H the scale height computed with the air temperature at an altitude of 1 km above the surface of Gale crater. GCM computations show that with MY 34, 35 and *clim* dust scenarios, we have to first order $e^{-\frac{\Delta z}{H(t_{Y_1})}} \sim e^{-\frac{\Delta z}{H(t_{Y_2})}}$. Thus Equation 5 leads to:

$$\frac{\langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle}{\langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle} = \frac{\langle P_{\text{MSL}}(t_{Y_1}) \rangle}{\langle P_{\text{MSL}}(t_{Y_2}) \rangle} = \beta \quad (6)$$

where β is by definition the interannual variability between the two years of measurements. Hence, as we only use a ratio of pressures, the absolute pressure values measured by MSL do not impact the absolute values of InSight pressure measurements after being corrected, and thus do not introduce a bias in our comparison. The problem described by Equation 6 can be transformed into the following optimization problem:

Find E that minimizes :

$$\| \langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle - \beta \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle \| \quad (7)$$

where $\| \cdot \|$ refers to the Euclidean norm. Introducing Equation 4 into 6 gives:

$$\begin{aligned} \langle P_{\text{InSight,Measured}}(t_{Y_1}) \rangle &= -\beta \langle P_{\text{InSight,Measured}}(t_{Y_2}) \rangle \\ &= \beta \langle E(T(t_{Y_2})) \rangle - \langle E(T(t_{Y_1})) \rangle \end{aligned} \quad (8)$$

We further assume that E can be written as a polynomial function of the sensor temperature:

$$E(T(t)) = \sum_{k=0}^n \alpha_k T(t)^k \quad (9)$$

Introducing this into Equation 8 finally leads to:

$$\begin{aligned} \langle P_{\text{InSight,Measured}}(t_{Y_1}) \rangle &= -\beta \langle P_{\text{InSight,Measured}}(t_{Y_2}) \rangle \\ &= \sum_{k=0}^n \alpha_k \langle \beta T(t_{Y_2})^k - T(t_{Y_1})^k \rangle \end{aligned} \quad (10)$$

This last equation represents a least-mean-square problem that can be solved numerically to determine the coefficients α_k of E for a given degree n . However, the problem must be constrained to have a physical solution. The first term α_0 is indeed poorly constrained as $\beta \sim 1$ (β ranges from 0.992 to 0.998 during the period considered below) and thus $\langle \beta T(t_{Y_2})^0 - T(t_{Y_1})^0 \rangle \sim 0$. A close look at Figure 2 reveals an unexpected increase of the uncorrected pressure at $L_s \sim 63^\circ$ (suggesting that E should be negative), and then a drop (suggesting that E should be positive), both certainly resulting from a rise of temperature at $T = 270$ K and followed by a decrease of temperature at $T = 275$ K. Such observations are also found at $L_s = 120^\circ$ and 290° , suggesting a change of behavior of the sensor temperature sensitivity, that is, a change in the sign of $E(T)$ close to $T = 273$ K (with $E(T > 273 \text{ K}) > 0$ and $E(T < 273 \text{ K}) < 0$). Hence we simply assume that:

$$E(T = 273 \text{ K}) = 0 \text{ Pa} \quad (11)$$

The resolution of the problem is made as follows. For each degree n , we compute the coefficients α_k with a least mean square algorithm based on Equations 10 and 11 to have E . We then compute $\| \langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle - \beta \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle \|$ using Equation 4. We iterate on the degree n to find which E is solution of the optimization problem described in Equation 7.

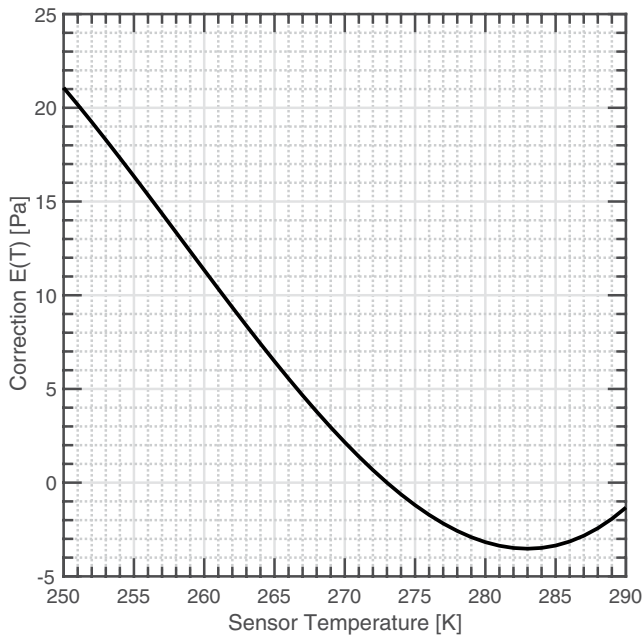


Figure 3. Correction of the thermal sensitivity E applied on the pressure measurements versus sensor temperature T .

To compute the least-mean square inversion, we use the data acquired at the end of the MY 34, at $L_s > 340^\circ$ to remove the effect of local dust storms, and data at $L_s < 21^\circ$ (the limit of the MSL data set that we used). We find that increasing the degree of E reduces the norm of $\langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle - \beta \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle$. However, polynomial corrections with high degree n induce non-physical behavior of the corrections, especially for the edges of the sensor temperature range (e.g., $T \sim 250$ K). We complete thus our test by investigating which degree minimizes the absolute difference $|\langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle - \beta \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle|$ so that our correction will give a realistic solution for the complete range of sensor temperature correction. With these two criteria, we find that $n = 3$ is the solution that reduces the euclidean norm and minimizes the absolute value of $\langle P_{\text{InSight,Corrected}}(t_{Y_1}) \rangle - \beta \langle P_{\text{InSight,Corrected}}(t_{Y_2}) \rangle$. This fit has also been validated with some goodness of fit criteria ($R^2 \sim 0.93$, reduced $\chi^2 \sim 0.94$). The final correction E (in Pa) can be at least written as a function of the sensor temperature T (in K):

$$E(T) \approx 5.5273 \times 10^{-4} T^3 - 0.4284 T^2 + 109.6849 T - 9.2602 \times 10^3 \quad (12)$$

This correction is presented in Figure 3. This correction add ± 5 Pa to the pressure measurements in the range of the sensor temperatures measured most often during the mission (Figure 2b). The correction is significant at very low temperatures (e.g., $T < 260$ K), as expected when looking at the drop of the surface pressure measured at the end of the mission. Finally, we logically find that $E(T < 273 \text{ K}) < 0$ and $E(T > 273 \text{ K}) > 0$.

Applying this correction to the complete measured pressure data with Equation 4 leads to the result presented in Figure 4. As expected, this correction strongly modifies the pressure measured by InSight in terms of amplitude and shape. The Northern winter surface pressure during InSight Year 1 is lower than during InSight Year 2, but tends to equalize during spring. These results are thus consistent with the analysis of contemporary MSL data

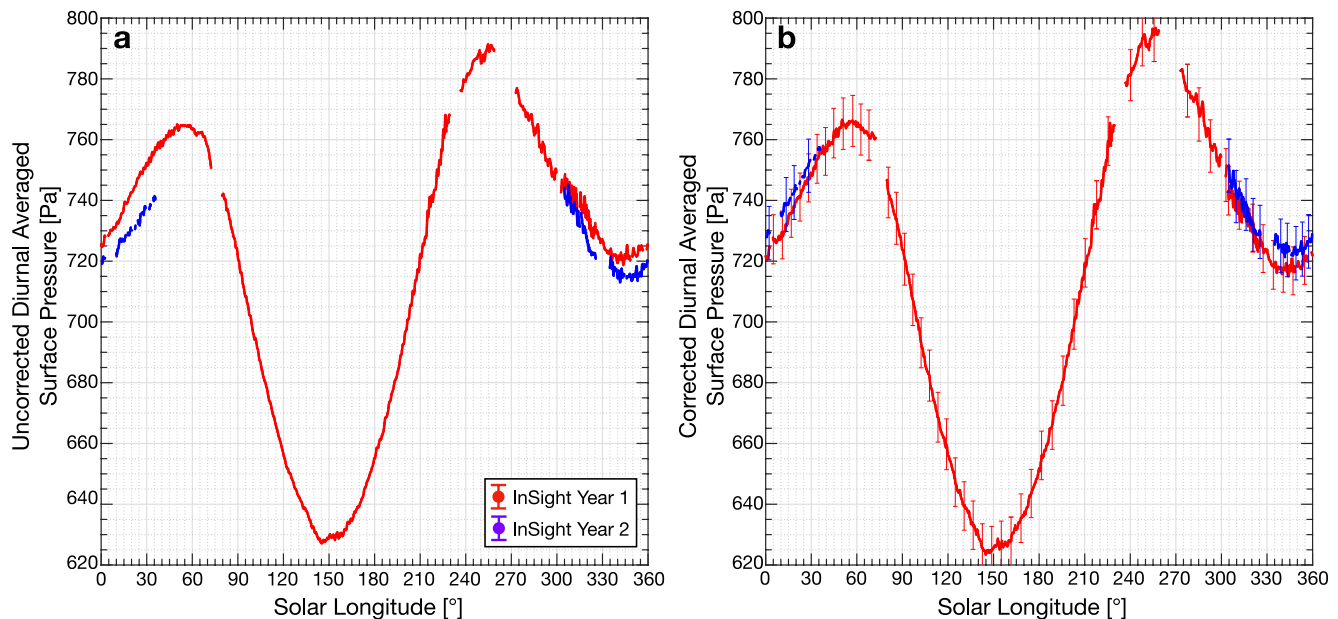


Figure 4. (a) Diurnal averaged surface pressure computed from the raw pressure data. (b) Diurnal averaged surface pressure after applying the thermal correction. Error bars represent the uncertainty on the measurements after the correction at $3 - \sigma$. The details of the uncertainty computations are described in the text. Red dots are for the first year of the mission, while blue dots are for the second year.

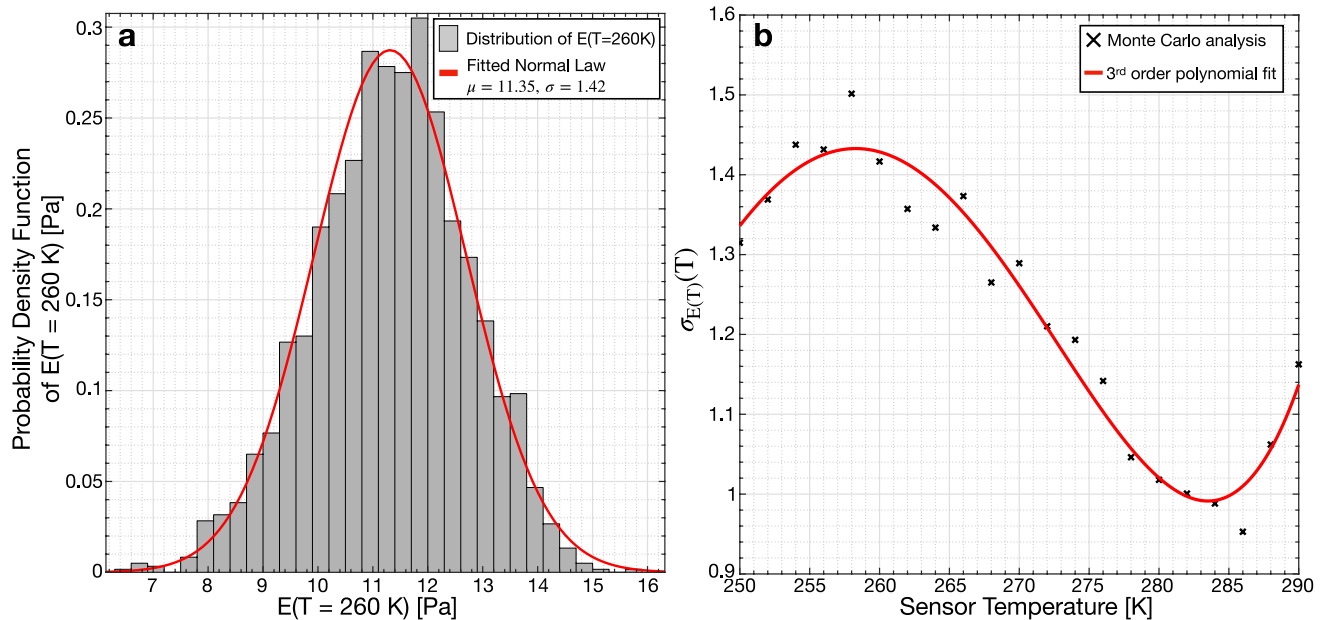


Figure 5. (a) Monte Carlo analysis to retrieve $\sigma_{E(T=260\text{ K})}$. The histogram of the samples is presented in gray and is normalized to obtain a probability density function. The fitted normal law is illustrated in red and has as parameters the mean μ and the standard deviation of the distribution σ . (b) Empirical law for $\sigma_{E(T)}(T)$ obtained from Monte Carlo analysis (black cross) and third order polynomial fit of this law (red line).

from de la Torre Juarez et al. (2019) who also observed a pressure deficit at the end of MY 34 and the beginning of MY 35 when studying the interannual variability of MSL pressure data.

3.4. Uncertainty of the Corrected Data

The accuracy of InSight pressure measurements is crucial for the determination of possible secular pressure changes and will be useful to the community for future scientific work. We thus need to quantify the uncertainty of the InSight corrected pressure proposed in Section 3.3. Three kinds of uncertainty can be highlighted here according to Equation 4. The first one is the uncertainty of the pressure sensor on the measurement, which is $\sigma_{P_{\text{sensor}}} = 1.5$ Pa R.M.S (Banfield et al., 2019). The second one is caused by the uncertainty of the correction. This correction deals with the uncertainties at $1 - \sigma$ of the pressure measured by InSight $\sigma_{P_{\text{sensor}}}$ and the sensor temperature uncertainty σ_T set here at 1 K. However, the uncertainty of β , and thus the impact of MSL uncertainty on our correction, is negligible. ($\frac{\sigma_\beta}{\beta} \ll 1$, Appendix A). Hence, as we use a ratio of MSL pressure measurements to derive the interannual variability β of the InSight pressure data, the absolute MSL pressure accuracy does not impact the accuracy of our correction for the InSight pressure measurements. Finally, an uncertainty is associated with the choice of the temperature nullifying the correction term (Equation 11); this last one having been made arbitrarily after analysis of the correlations between the measured pressure and the sensor temperature.

To derive the uncertainty in E at a sensor temperature T ($\Delta E(T)$), we perform a Monte Carlo error analysis as described in Press et al. (1993) or Forget et al. (2007). We generate an ensemble of 10^4 inputs (P_{InSight}, T), affected by the various uncertainties described above. All the input parameters are computed using their nominal values plus random values computed from a normal distribution with a standard deviation associated with $\sigma_{P_{\text{sensor}}}, \sigma_T$. The condition provided in Equation 11 is also perturbed using a normal distribution with a standard deviation of σ_T . We then apply our algorithm to retrieve the thermal correction E at a given sensor temperature T with these inputs. We finally compute the standard deviation of the E provided. We find that the distribution of the E retrieved follows a normal distribution as illustrated in Figure 5a. The standard deviation of the fitted normal distribution gives the uncertainty of $E(T)$ at $1 - \sigma$. We apply this Monte Carlo analysis for temperatures ranging from 250 to 290 K to retrieve $\sigma_{E(T)}$, that is, the uncertainty at $1 - \sigma$ level. The results from this computation are presented in Figure 5b. The variations of this curve follow the variations of the gradient of $E(T)$. We then do a least mean square polynomial fit to have an empirical law to simply deduce $\sigma_{E(T)}(T)$:

$$\sigma_{E(T)}(T) = 5.1453 \times 10^{-5} T^3 - 0.0418 T^2 + 11.2738 \times T - 1.0109 \times 10^3 \quad (13)$$

We finally retrieve the $3 - \sigma$ uncertainty of one pressure measurement by combining these two uncertainties, plus a term due to the dependence between the measurement and the thermal correction, as the raw measurements and temperature are correlated due to the initial calibration procedure:

$$\Delta P_{InSight}(T) = 3 \times \sqrt{(\sigma_{P_{sensor}})^2 + (\sigma_{E(T)}(T))^2 + 2\sigma_{P_{sensor}}\sigma_{E(T)}(T)\rho_{P_{sensor},T}} \quad (14)$$

where $\rho_{P_{sensor},T}$ is the correlation coefficient between the raw pressure measurement P_{sensor} and the sensor temperature (T), assumed to be 1 as the pressure sensor is calibrated using the sensor temperature (Banfield et al., 2019). Uncertainties on the corrected pressure range from 7.5 to 8.9 Pa at a $3 - \sigma$ level. Such values are close to the magnitude of the atmospheric mass variations expected based on Thomas et al. (2016) (± 9 Pa difference between Viking 1 and InSight surface pressures), but are much smaller than the expected changes that are computed from Blackburn et al. (2010); Malin, Bell, et al. (2001) ($\sim +25$ Pa difference between Viking 1 and InSight surface pressure). We therefore consider that the InSight corrected pressure data are accurate enough to detect such secular pressure changes.

3.5. Comparison With MSL Pressure Data and Validation

To test the reliability of our correction, we propose here to compare the corrected InSight pressure to the MSL pressure measurements interpolated to the InSight landing site. This comparison is relevant as the use of the MSL data to correct the InSight data relied on the year-to-year ratio (Equation 6), and thus does not influence seasonal variation given by InSight pressure data after the correction. To do so, we use the methodology described in 2.2 by using the ratio $\frac{P_{MSL}}{P_{GCM,MSL}}$ into Equation 3, with MY 34, 35 and *clim* dust scenario for the beginning of MY 36. The comparison between interpolated MSL pressure and InSight measurements is presented in Figure 6. There is an overall good agreement between the InSight corrected measurements and the MSL pressure measurements. This consistency strengthens the credibility of our correction.

We note a deficit δ of pressure between MSL pressure interpolated at the InSight landing site, and InSight corrected pressure between $L_s \sim 220^\circ$ and $L_s < \sim 360^\circ$ (Figure 6). This deficit represents $\sim 1\%$ of InSight surface pressure and could reach 8–10 Pa. Three causes could explain this deficit: 1) a dynamical effect i.e., not included in our interpolation process 2) a meteorological effect that changes the thermal state of the atmosphere, and thus the scale height used during the interpolation 3) a problem with our correction.

As InSight and MSL are close to each other (~ 600 km), the impact of atmospheric dynamics on the interpolation is limited. As underlined by Figure 6b, the deficit δ remains when correcting the large-scale pressure gradients. Therefore, these large-scale atmospheric dynamics do not explain the pressure deficit δ . Another possible explanation might be the small-scale/regional topography of Gale crater that is not included by our GCM. The two closest points are at a longitude of 135°E and 140.6°E , with an altitude of -2069 and -1879 m respectively. The interpolation using the four closest points of the GCM to the MSL landing site gives an altitude of -1544 m, far from the actual landing site altitude of $-4,501$ m. Furthermore, complex crater circulations (Pla-Garcia et al., 2016; Rafkin et al., 2016) might impact the pressure measured by MSL. To investigate the potential impact of this particular topography (and thus the local circulations, aerosol distributions, small-scale physical phenomena, etc. (Spiga & Forget, 2009)) on the surface pressure deficit, we used the mesoscale LMD model simulations described in Section 2.1. We ran the model for 24 hr after initial spin-up time of 24 hr, at $L_s = 270^\circ$. The mesoscale model, resolving more accurate topography, helps to reduce the pressure deficit between MSL interpolated to InSight and InSight measurements by 2–3 Pa. However, it still does not fully explain the difference observed. Hence, the deficit δ does not seem to be due entirely to local dynamic effects, or to the too coarse resolution of the GCM which does not capture the topography of Gale crater.

We then studied this deficit δ by investigating the possible influence of the scale height H , using the interpolation described in 2.1. Results are presented in Figure 7a. We observe again the pressure deficit between InSight and MSL after $L_s = 180^\circ$. To study the influence of the scale height, we compute the temperature T_* such that:

$$\frac{<P_{MSL \rightarrow InSight}>}{<P_{InSight}>} \approx 1.$$

Using Equations 1 and 2, T_* writes:

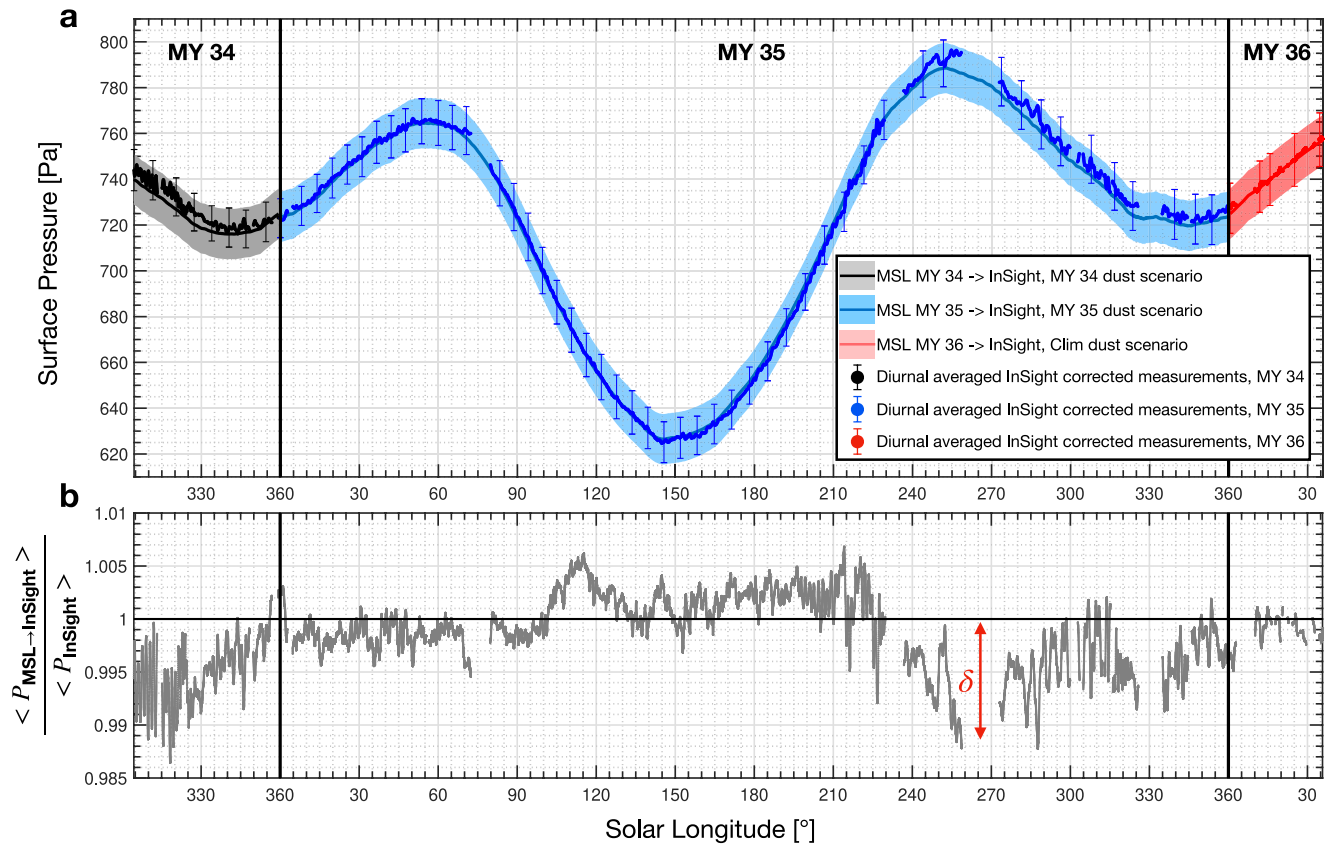


Figure 6. (a) Comparison between the surface pressure measured by InSight and that measured by MSL but interpolated to the InSight landing site, for MY 34, 35, and 36. The filled box around the plain line depicts the $3 - \sigma$ uncertainty of the interpolation due to weather-induced uncertainty and MSL absolute errors, following the methodology presented in Section 2.2. Pressure interpolated is averaged over a period of 15 sols to remove atmospheric tides and baroclinic activity. InSight measurements are diurnally averaged thus still indicate baroclinic activity with periods of several sols. The error bars correspond to the $3 - \sigma$ on InSight corrected pressure measurements as described in Section 3.4. (b) Evolution of the ratio of MSL REMS pressure measurements interpolated to the InSight landing site, and InSight pressure measurements. Dots correspond to the ratio using the interpolation method described in Section 2.2.

$$T_* = - \frac{\Delta z}{\frac{R}{\mu g} \ln \frac{P_{\text{MSL, measured}}}{P_{\text{InSight, measured}}}} \quad (15)$$

where Δz correspond to the difference of altitude between the InSight landing site and MSL altitude (in meters), and $P_{\text{MSL, measured}}$ is the raw MSL REMS pressure measurements. To detect any anomaly in the temperature, we compare this temperature T_* to the temperature at an altitude of 1 km above the surface predicted by the GCM (Figure 7b). This comparison underlines warming by 10–15 K of the temperature at this altitude, at $200^\circ < L_s < 360^\circ$.

This difference between T_* and the temperature given by the GCM might be explained by an unexpected accumulation of aerosols within Gale Crater, such as dust, compared to what is assumed in the GCM. The presence of aerosols would indeed warm up the air as they absorb solar radiation. Moreover, GCM simulations using our *warm* scenario indicate a warming of nearly 5–8 K of this atmospheric layer, which is the order of magnitude of the anomaly observed here. Hence, by studying the evolution of the temperature anomaly presented in Figure 7b, we could assume that at $L_s > 180^\circ$, there are local effects that increase the quantity of dust or other aerosols within the crater, inducing a warming of the air temperature. By comparing the measured and interpolated pressures at different local times around $L_s = 275^\circ$ (Figure 7c), we find that this pressure anomaly is much more significant during the daytime periods than at night. We can assume that the air within the crater is heated during the day due to the presence of this dust in suspension. This hypothesis is consistent with the REMS temperature observations

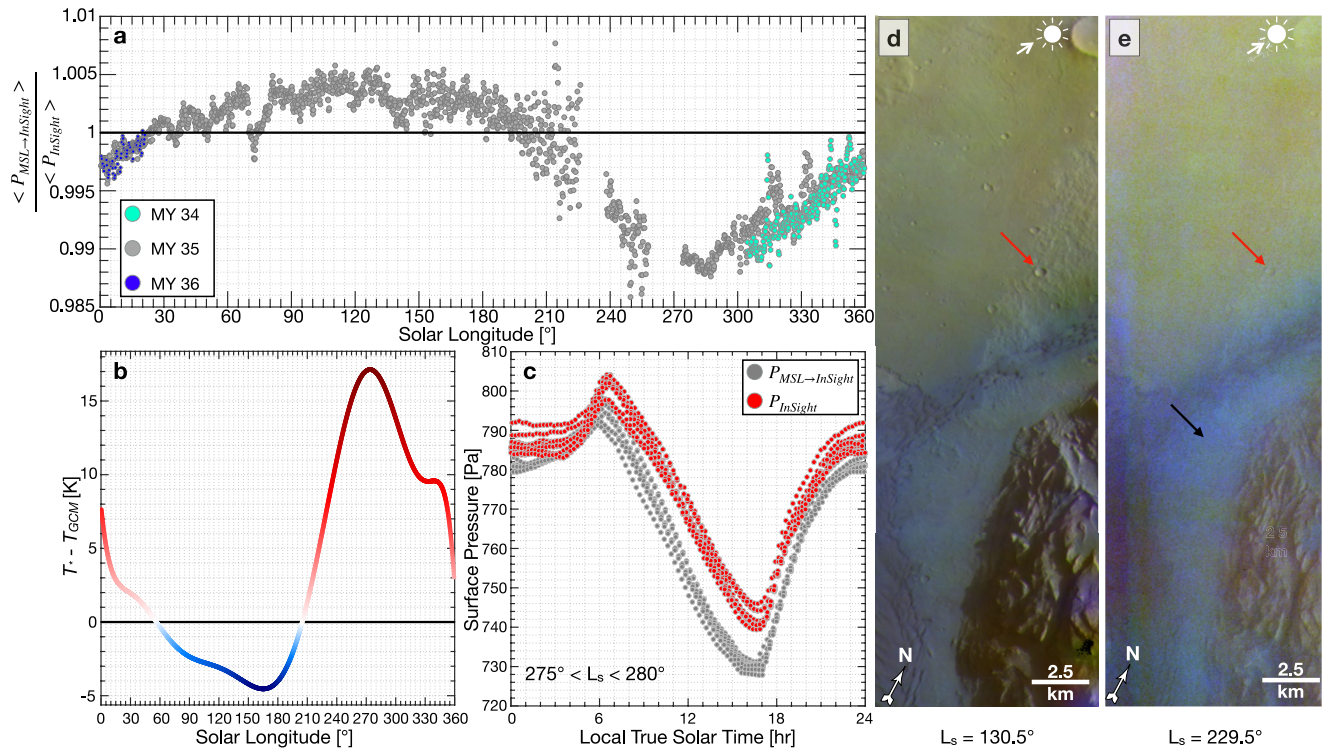


Figure 7. (a) Evolution of the ratio of MSL REMS pressure measurements interpolated to the InSight landing site, and InSight pressure measurements. Dots correspond to the ratio using the interpolation method described in Section 2.1, that is, neglecting atmospheric dynamic effects, during MY 34 (green), MY 35 (gray), and MY 36 (blue). (b) Anomaly between the temperature of the GCM at an altitude of 1 km above the surface, and the temperature T_s that gives a ratio of 1, as a function of L_s (colored curve) for MY 35. (c) Comparison between InSight surface pressure over a complete sol and MSL pressure interpolated at InSight landing site between $L_s = 275^\circ$ and 280° , during MY 35. (d) Extract of THEMIS image V63417011 of Gale Crater (center of the original image: 4.9°S ; 137.0°E) taken at $L_s = 130^\circ$, LTST = 7.2 hr, with a solar incident angle of 74.5° . (e) Extract of THEMIS image V65575024 at the same location, taken at $L_s = 229^\circ$, LTST = 7.2 hr, with a solar incident angle of 71.3° . The black arrow on (e) points to the suspected aerosols, whereas the red arrows on (d) and (e) point to the same crater for a comparison of the perceptibility of the ground. White arrows point to the position of the Sun in the sky.

at the surface and at 2 m from the ground at this time of year, which are respectively lower and higher than predicted by mesoscale models (Pla-Garcia et al., 2016).

Assuming that Gale crater is full of dust at this time of the year is plausible as several observations have shown the large presence of dust within the crater at this time of the martian year. Measurements of the line-of-sight across-crater extinction with MSL cameras report an increase of dust loading at lower elevations during the dusty season, that is, $180^\circ < L_s < 360^\circ$. The analysis of UV sensors data onboard MSL also confirms this observation, with net dust lifting from the crater floor during the dusty season, and net deposition during the rest of the year (Vicente-Retortillo et al., 2018). Such observations confirmed models of dust diffusion rate within Gale crater (Moore et al., 2019) that report net dust lifting from $L_s \approx 220$ – 240° , and net dust deposition in the crater before this date, explaining the variation of the thermal inertia of the ground (Rangarajan & Ghosh, 2020). This behavior of settling and suspension of dust within Gale might be explained by the dynamics of the planetary boundary layer within the crater. Fonseca et al. (2018) points out that at $L_s > 180^\circ$, the planetary boundary layer (PBL) height is higher than the crater rim for a few hours during the afternoon, inducing a mixing between the air outside and inside the crater. In addition, dust might be injected within the crater because of dust devils and wind-driven dust lifting. As reported by Steakley and Murphy (2016); Kahanpää et al. (2016); Ordonez-Etxeberria et al. (2018); Newman et al. (2019), there are very strong seasonal variations in dust devil activity, with a peak of that dust devil lifting around southern spring and summer, that is, when we observed the pressure deficit δ . Most of the dust devil occur during the day, with a peak of activity around noon, close to the period of the sol when the pressure difference between MSL and InSight is the most important (Figure 7c).

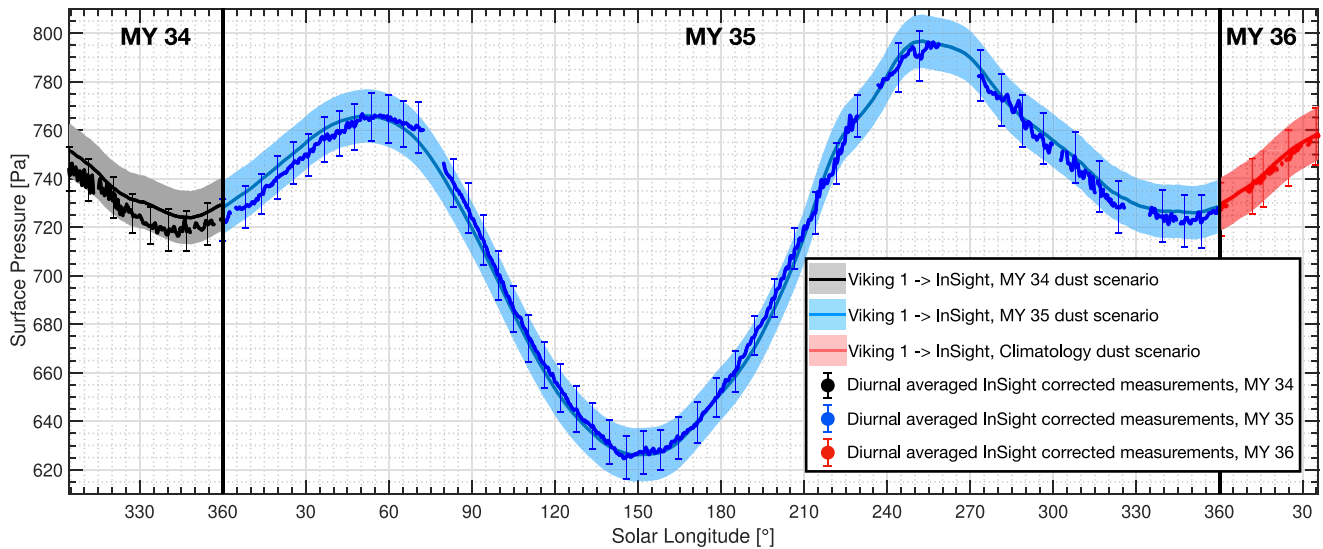


Figure 8. Comparison between the surface pressure by Viking 1 interpolated at the InSight landing site for MY 34, 35, and 36. The filled box around the plain line depicts the $3 - \sigma$ uncertainty of the interpolation detailed in Section 2.2. Pressure interpolated is averaged on a period of 15° to remove atmospheric tides and baroclinic activity. InSight measurements are diurnal averaged and still keep baroclinic activity. The error bars correspond to the $3 - \sigma$ on InSight corrected pressure measurements as described in Section 3.4.

We also obtain indications of the presence of aerosols near the surface using the THEMIS visible camera (Christensen et al., 2004). Figures 7d and 7e compare two images of Gale Crater, at the same local hour, in quasi-similar illumination conditions, but at two different L_s (130° and 229° respectively). Figure 7e clearly shows the presence of aerosols (black arrow) confined within the crater, as the portion of Mount Sharp remains easily detectable and less obstructed. Another indication of the presence of a significant quantity of aerosols in the air is the difficulty of detecting the ground and the craters at the bottom of Gale crater on the image e compared to d (see red arrows), in quasi-similar illumination conditions. Water fog is a suspected candidate to explain this phenomenon, as the image is taken during the early morning, but it seems highly unrealistic as the relative humidity at this time of the year is at its lowest level (Martínez et al., 2016, 2017). Furthermore, images taken at nearly 17 hr Local True Solar Time (LTST) also report such features (see for instance THEMIS image V59356002 that were taken at a quite similar location at $L_s = 335^\circ$, at 17.37 hr). However, opacity derived from MSL cameras does not show a significant increase of dust loading at this time, compared to what is predicted by the *clim* scenario (see Figure 9 of Ordóñez-Etxeberria et al. (2019)).

We acknowledge that these observations are consistent with the presence of dust within the crater, but do not show that Gale crater is abnormally dusty at this time of the year. The Northern fall and winter are indeed expected to be dustier than the beginning of the martian year (Kahre et al., 2017), as previously observed at the beginning of MSL mission (Guzewich et al., 2017). Further investigations like Moores et al. (2015); Guzewich et al. (2017) using MY 35 data will help to validate or deny our assumption of an abnormally dusty Gale crater at this time of the year. Nevertheless, our hypothesis is still credible because of the large presence of dust within the crater at the time of the year, and has the potential to explain the observed deficit. A major mistake in our correction seems at least unlikely because of the very good agreement between the pressure measurements of MSL and InSight during the rest of the year.

4. Results: Comparison With Viking Lander 1 Pressure Data

The comparison between the Viking 1 surface pressure measurements interpolated at the InSight landing site and the InSight temperature-corrected measurements for MY 34, 35, and 36 is presented in Figure 8. During MY 34 and the beginning of MY 35 ($L_s < 55^\circ$), InSight pressure measurements are lower compared to Viking 1 pressure by 5–10 Pa. de la Torre Juárez et al. (2019) also reported a pressure deficit at these times when studying the repeatability of MSL pressure data. Using MCS thermal data, they relate this to a possible increase of the NSPC expansion during MY 34 compared to MY 33. Such an expansion would consequently decrease the

atmospheric mass at this time, reduce the surface pressure, and thereby explain the deficit observed. This deficit is not observed during Northern winter of MY 35 with MSL and InSight pressure data, and thus cannot be linked to a secular change.

After the sublimation of the NSPC during MY 35, the InSight pressure measurements match Viking pressures very well within the uncertainties associated with the interpolation method. The weather-induced uncertainty might explain the small deficit of pressure ($\sim 2\text{--}3$ Pa) observed at $L_s > 250^\circ$ because Viking 1 pressure was more affected by baroclinic activity as the lander is at a higher latitude than InSight. There is also less confidence in the Viking 1 pressure average during this period, as a lot of the measurements available at this time of the year were affected by the first global dust storm recorded by Viking (Ryan & Sharman, 1981), and thus removed from the data set.

However, the overall strong agreement between Viking 1 interpolated surface pressure and InSight thermally corrected measurements strongly supports the assumption that the atmospheric mass has not changed since the Viking era, nearly 40 Earth years before the InSight era. More precisely, the comparison between InSight and Viking 1 pressure data suggests that the atmospheric mass has not changed by more than ± 8 Pa, knowing that our method has an accuracy of ~ 11 Pa at $3 - \sigma$. Our results suggest that SPRC mass balances from Malin, Bell, et al. (2001); Blackburn et al. (2010) might have been overestimated, but support low estimated values of atmospheric mass gain/loss due to the evolution of the SPRC (Thomas et al., 2016). Such results thus reinforce the assumption that the SPRC does not suffer from major changes over decades, as indicated by both imagery comparisons since the Mariner era and recent imagery data set (Piqueux & Christensen, 2008; Thomas et al., 2016). In fact, the SPRC might be varying with periods of erosion due to large summer dust events, followed by a period of deposition in the next winter (Becerra et al., 2015; Bonev et al., 2008; Byrne et al., 2015; James et al., 1992, 2010; Thomas et al., 2016). Further discussion on the role of dust events in the condensation and sublimation of CO_2 ice is shown in Section 5.2. The possible influence of CO_2 reservoirs under the SPRC on the durability of this cap has also been explored recently (Buhler et al., 2020).

The strong agreement of the surface pressure comparison during the formation and sublimation of seasonal caps (excluding the Northern winter of MY 34) also suggests a low variability of the martian seasonal CO_2 cap dynamic. The interannual variability of the recession of the seasonal polar cap has been widely studied in the literature (Benson & James, 2005; Brown et al., 2010, 2012; Calvin et al., 2015, 2017; Giuranna et al., 2007; James & Cantor, 2001; James & Lumme, 1982; James et al., 1979, 1992, 1996, 2000, 2001, 2010; Kieffer, 1979; Kieffer et al., 1972, 2000; Langevin et al., 2007; Piqueux et al., 2015). The recession of the seasonal caps has been monitored through the optical properties of the CO_2 ice caps and albedo contrast with neighboring surfaces by Viking orbiters (James, 1979, 1982; James et al., 1979; James & Lumme, 1982) and also completed with telescopic observations (James et al., 1992). Further observations in the 90s/2000s with the MOC camera on-board the Mars Global Surveyor orbiter (Benson & James, 2005; James & Cantor, 2001; James et al., 2000, 2001) and telescopic observations (James et al., 1996) have shown good repeatability in the recession curves, despite some local differences. This similarity in the retreat of the seasonal caps has also been confirmed with spectroscopic studies (Brown et al., 2010; Langevin et al., 2007) but their observations have limited coverage. More recently, Calvin et al. (2015); Calvin et al. (2017) have shown, using MARCI images taken recently (MY 28–31), that the retreat of the seasonal caps was globally similar to the ones observed by Viking, even if some discrepancies appear at discrete locations. However, exploiting these observations has been impeded by the challenging illumination conditions of the polar caps during the recession periods (see Piqueux et al. (2015) for a complete review of these issues). To counterbalance these limitations in the visible, the retreat of the seasonal caps has also been studied by thermal measurements in the infrared. The first observations were made by Viking (Kieffer, 1979; Kieffer et al., 1972) and the most recent studies (Giuranna et al., 2007; Kieffer et al., 2000; Piqueux et al., 2015) have again shown high repeatability in the retreat time-lapse of the seasonal ice caps. Small interannual variabilities exist but are mainly linked to the influence of global dust storms (Calvin et al., 2015, 2017; Piqueux et al., 2015) (see Section 5.2 for a discussion on this topic).

Here, the good agreement between the InSight and Viking surface pressure (Figure 8) during the formation and sublimation of seasonal caps during MY 35 also confirms the good repeatability in the recession of the polar caps between the Viking period and present. This result is contrary to a faster retreat of the SSPC that could be assumed when looking at Figure 2a. Indeed, the large difference in pressure observed in Figure 2a) (~ 20 Pa) would imply a lower extent of the SSPC when comparing SSPC observed during the Viking period and today. To

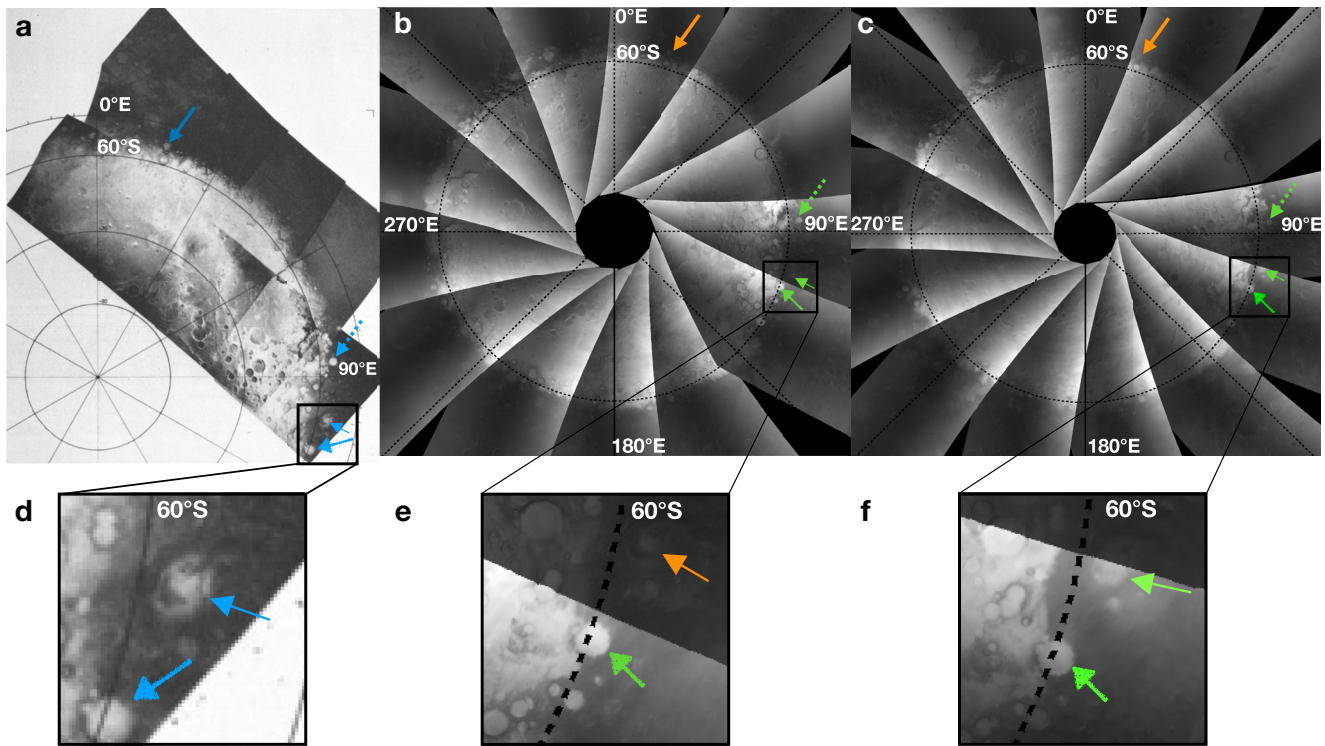


Figure 9. Comparison of the SSPC images taken by (a) Viking orbiter during MY 12, $L_s = 192.6^\circ$ (extracted from James et al. (1979)); (b) MARCI during MY 33, $L_s = 192.3^\circ$; and (c) MARCI during MY 35, $L_s = 192.9^\circ$. Blue arrows flag characteristic surface features for the comparison like craters. Orange arrows indicate a possible difference between the Viking images and MARCI images while green arrows indicate a good match between the images. (d) to (f) are zoom on the lowest flagged craters of (a), (b), (c). The 60°S circle of latitude on image (d) extracted from James et al. (1979) is misplaced, but arrows point to the same elements.

further refute the possible faster retreat of the SSPC (and thus validate again our correction), as well as the good agreement between Viking's observations of the SSPC retreat with today's observations, we propose in the following a comparison of the caps' albedo taken at these two periods. To do so, we exploit images of the SSPC taken by Viking Orbiter 2 during MY 12 (James et al., 1979) and MARCI images (Malin, Caplinger, & Davis, 2001) taken during MY 35 (images from MY 34 are not used because of the global dust storm that occurred during this year, hindering the visual detection of the caps). Piqueux et al. (2015) noted interannual variability in the caps' dynamics due to global dust storms. We thus add MY 33 to the comparison as a control year in case the global dust storm at the end of MY 34 influenced the cap dynamics during MY 35. Furthermore, even if the cap boundary is composed of water ice after the sublimation of the seasonal CO_2 ice (see the spectroscopic study in Langevin et al. (2007)), we assume that the albedo comparison between the Viking decade and late 2010s/early 2020s also reflects possible changes in the CO_2 cycle and the sublimation of the SSPC.

Details on the composition of MARCI polar mosaic are given in Calvin et al. (2015); Calvin et al. (2017). We select $L_s = 192.6^\circ$ for the comparison as Viking mosaic is available at this time (Figure 5 of James et al. (1979)) and as InSight uncorrected pressure data present a large difference with Viking 1 pressure data at this time of the year (Figure 2a). Similar analysis and conclusions can be drawn using other L_s (not shown here). The comparisons are presented in Figure 9. On Viking images, we flag with blue arrows craters or easily distinguishable topographies that are covered by ice at the boundary of the cap. We then look at MARCI images to see if the element is still covered by ice at the same time of the year. In this case, the element is flagged by a green arrow, whereas in the case of a divergence with Viking observations, the element is flagged by a red arrow. In case of doubt about the presence of ice, we flag the crater with an orange arrow. The comparisons (Figure 9) underline that no major changes have happened in the dynamic of the seasonal caps as the same extend of CO_2 ice is observed. It thus confirms what has been observed by comparing Viking 1 and InSight pressure data. The good agreement between these observations refutes the possibility of a discrepancy in atmospheric mass between Viking and InSight during the sublimation of the SSPC. It also reinforces the credibility of our correction, as the Viking and InSight pressure curves match well after the recalibration. Little variability can be noted as revealed by the orange arrows.

It can be explained by some observational biases. First, MARCI mosaics are built with images taken during all the day, and at different LTST. Hence, some ice might have sublimated during the day and would not be present on the mosaic. Second, the discrepancies on Figure 9f are actually a consequence of the timing of the mosaic, as the images are not taken exactly at the same L_s and illumination conditions. At least, it is very unlikely that these discrepancies observed are due to a faster retreat of the seasonal cap as a consequence of the MY 34 global dust storm. Our GCM simulations show indeed that this retreat should occur at the same speed between MY 33 and 35. Finally, we acknowledge that this comparison is limited as we are looking at specific topographic features to compare the SSPC observed by Viking and InSight. Future analysis using MARCI observations should make a more complete study of the SSPC boundaries (like Calvin et al. (2017) for instance) to definitively conclude on the good agreement between Viking 2 orbiter and MARCI images of the SSPC.

Thus, the comparison between Viking 1 and InSight pressure data, as well as the preliminary comparison of images taken by the Viking 2 orbiter and the MARCI camera during MY 35 suggest the absence of secular pressure changes or modifications in the dynamics of the seasonal ice caps.

5. Discussion

5.1. Evolution of the Atmospheric Mass Since MY 29

The non-detection of atmospheric mass changes between the 1970s and present disagrees with the conclusions obtained from the comparison between Viking and Phoenix surface pressure made in (Haberle & Kahre, 2010). The preliminary comparison between MSL and Viking 2 pressure data, which are nearly at the same altitude above the surface, did not show any significant increase of the atmospheric mass, but rather possibly a small decrease. We propose to extend this analysis by also comparing Phoenix and MSL data with Viking 1 pressures, using our methodology presented in Section 5.1 for the comparison between Viking 1 and InSight pressure. Phoenix data used here are extracted from Taylor et al. (2010), as they are thus corrected from the temperature gradient within the sensor that disturbed the measurements (Taylor et al., 2010). The results are presented in Figure 10.

Figure 10a underlines an excess of pressure when comparing Phoenix measurements to Viking 1 interpolated to Phoenix landing site. Such result is consistent with the analysis from Haberle and Kahre (2010). However, the location of Phoenix must be considered and might qualify their conclusions. Phoenix landed at a high latitude (68.22°N) compared to the other measurements used in this study that were made at mid/equatorial latitudes. This difference could lead to errors in our interpolation due to the large latitudinal pressure gradients. The error bars in Figure 10a underlines that the difference lies within the $3 - \sigma$ uncertainty of our interpolation method. Hence, it is difficult to conclude on a possible increase of atmospheric mass in MY 29 using Phoenix measurements only. These observations might actually illustrate a real rise of the atmospheric mass due to a significant SPRC erosion during the MY 28 global dust storm (Montabone et al., 2015). Bonev et al. (2008); Becerra et al. (2015); Byrne et al. (2015); Thomas et al. (2016) report that southern spring/summer dust storms, like the one in MY 28, might enhance the SPRC sublimation, which would consequently increase the atmospheric mass. Further observations like an analysis of the SPRC extent between MY 27, 28, 29 should help to find the explanation of this increase of the surface pressure at the Phoenix landing site.

The comparisons of MSL and InSight data with Viking 1 pressure measurements both show the same results, that is, an excess of pressure for $90^\circ < L_s < 180^\circ$, and a deficit elsewhere. Such divergences are small (less than 1% generally) and both comparisons are consistent, that is, MSL and InSight present a deficit or excess of pressure at the same time of the year, but MSL data have sometimes larger relative differences. We note a repeatable annual cycle in Figure 10a but this is actually a small systematic and seasonal error in the estimation by the GCM of the meteorological component of the pressure cycle (Hourdin et al., 1993). We study with Figures 10b–10d the evolution of these divergences to Viking interpolated pressure, at three times of the year: $L_s = 20^\circ, 170^\circ, 310^\circ$. No clear trend can be established when comparing MSL or InSight to Viking 1 pressure data, thus rejecting the idea of a monotonic SPRC mass balance over the years. The possibility of an erosion of the SPRC following the storm in MY 34 is difficult to show from the pressure data. First, the pressure recorded by MSL at the end of MY 34 was strongly impacted by a local dust storm (Viúdez-Moreiras et al., 2020). Moreover, when the storm stopped, the northern seasonal cap was still forming, with potentially an anomalous extent that lead to a decrease in the available atmospheric mass (de la Torre Juárez et al., 2019). The uncertainty in both the data and the

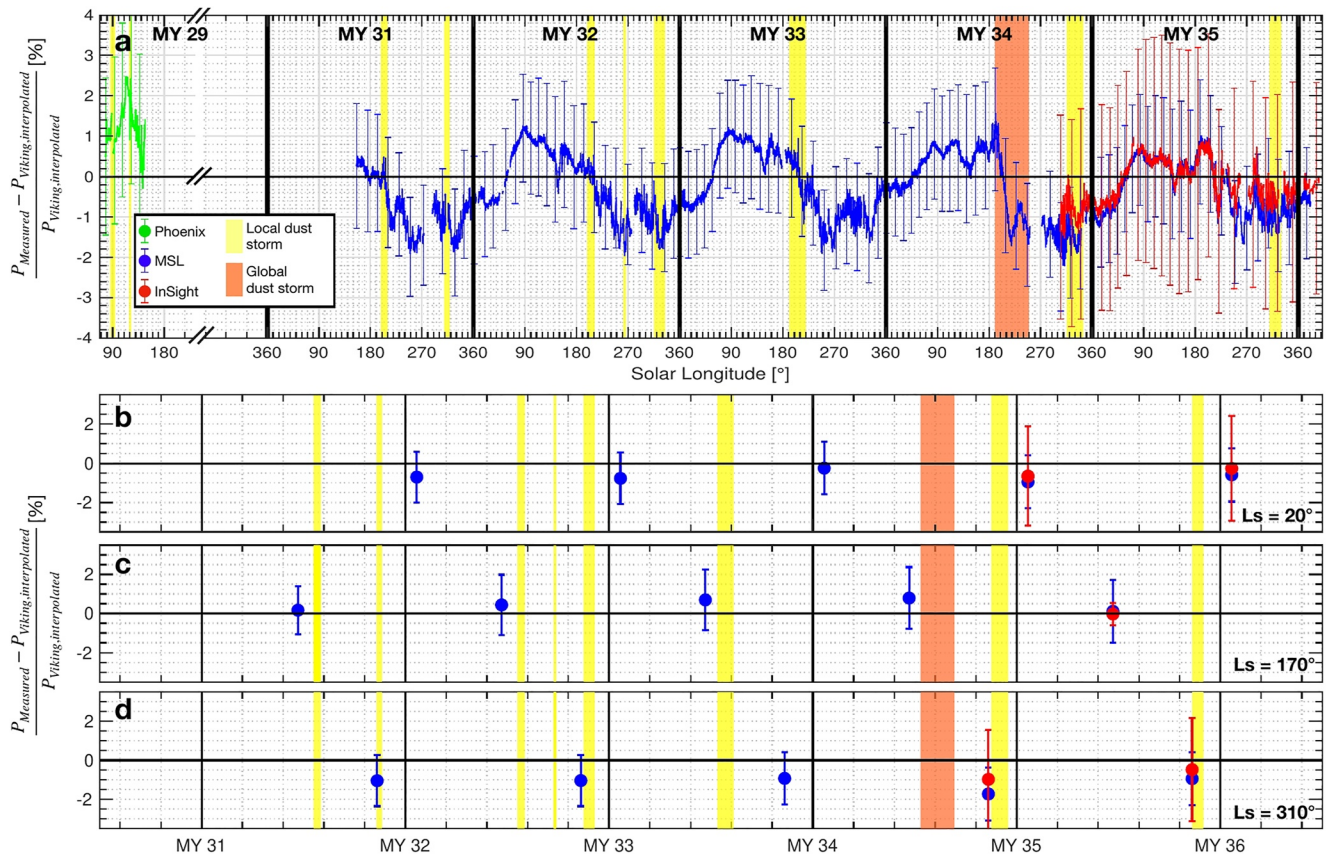


Figure 10. (a) Comparison of the surface pressure measured by Phoenix (green dots), MSL (blue dots), and InSight (red dots), to Viking 1 measurements (interpolated at each landing sites), from MY 29 to MY 36. Yellow boxes correspond to periods of local dust storms at landing sites (Holstein-Rathlou et al., 2010; Ordonez-Etxeberria et al., 2019), while the orange box corresponds to the period of MY 34 global dust storm (Lemmon et al., 2019; Viúdez-Moreiras et al., 2019). (b) to (d) Evolution of the relative difference between Viking 1 interpolated and MSL (blue) and InSight (red), as a function of martian year at $L_s = 20^\circ$ (b), $L_s = 170^\circ$ (c), $L_s = 310^\circ$ (d). The error bars indicate the sensitivity of the comparison with regards to the interpolation uncertainty at $3 - \sigma$, as described in Section 2.2.

interpolation, represented by the error bars in Figure 10, also explain the deficit observed by InSight and MSL. These two comparisons suggest again that there is no significant long-term pressure change.

5.2. CO₂ Cycle and Dust

What can induce year-to-year variations in the seasonal CO₂ ice budget? As reviewed in Titus et al. (2017), the CO₂ ice condensation and sublimation rates are controlled by the local energy balance, as the CO₂ condenses or sublimates in the exact amount needed to keep the surface and atmosphere at the CO₂ condensation temperature when ice is present.

At a given season, this energy balance could fluctuate from one year to the other. This stems from interannual changes in both CO₂ ice albedo and emissivity, as well as changes in the incident infrared radiation due to variations in the heat advected by the atmosphere or by the clouds. It is also sensitive to the amount of heat stored in the subsurface during previous seasons: the heat conducted from the subsurface up to the CO₂ ice on the surface depends on the subsurface temperatures, which are themselves influenced by the temperature from the previous summer when no CO₂ ice was present.

On these grounds, atmospheric dust can influence the CO₂ budget in a variety of ways:

First, during the condensation phase (i.e., in the polar night), dust primarily increases the thermal emissivity of the atmosphere and thus its radiative cooling (Pollack et al., 1990). More CO₂ condenses in the atmosphere and less on the surface. The net effect is an observed decrease of the thermal infrared emission at the top of the atmosphere due to the radiative effect of CO₂ clouds and/or the lower emissivity of the CO₂ snow freshly deposited

from the atmosphere (Cornwall & Titus, 2009; Forget & Pollack, 1996). This means less CO₂ ice condensing during a dust storm reaching the polar night. CO₂ ice deposits that condensed in the presence of extra dust may also be durably modified. They could have a higher albedo because they were formed from larger fractions of small particles condensed in the atmosphere, but their albedo could also be lowered by the contamination of more dust particles. Which effects dominate? Looking at the seasonal deposits around the north pole, Byrne et al. (2008) found that the northern fall 2001 global dust-storm resulted in slightly brighter ice deposits in the following spring. They considered this result to be “counter-intuitive.” It can probably be attributed to comparatively more atmospheric condensation in fall enhancing the spring albedo. The amount of airborne dust also influences the atmospheric circulation and thus the transport of heat and the dilution of non-condensable gas, these last ones influencing the CO₂ condensation temperature (Forget et al., 2008; Piqueux et al., 2020). Airborne dust can impact the structure of the polar vortices (Ball et al., 2021; Guzewich et al., 2016; Streeter et al., 2021), possibly inducing a warming of the Northern polar vortex (Guzewich et al., 2016) that can affect the CO₂ condensation rate (Zhao et al., 2021). Conversely, the acceleration of the meridional wind speed induced by an increase of dust loading can lead to an acceleration of the northern CO₂ condensation process (Zhao et al., 2021). All of these possible modifications of the atmospheric dynamics and their impact on the CO₂ ice budget are however sensitive to the timing of the dust loading (Zhao et al., 2021).

Second, during the sublimation phase, or more generally when CO₂ ice is significantly sunlit, the net effect of airborne dust is also equivocal, as studied by Bonev (2002); Bonev et al. (2008). Airborne dust redistributes the downward radiation from solar to thermal infrared because dust absorbs solar radiations and re-emits at thermal wavelengths. Model calculation and camera observations show that regions of high-albedo CO₂ frost will sublimate faster with more airborne dust (as they mostly absorb in the thermal range) whereas low-albedo regions will sublimate slower (as they mostly absorb in the visible) (Bonev et al., 2003, 2008).

Third, during summer (when no CO₂ ice is present) airborne dust could also modify the mean surface temperature at high latitude and the stored subsurface heat, but once again the net effect is subtle and depends on the atmospheric temperatures and surface albedo.

Overall, determining the net effect of regional and global dust storms on the seasonal CO₂ cycle is not straightforward as the different processes involved could tend to balance each other. This may explain why the seasonal cycle was observed to be relatively insensitive to the occurrence or non-occurrence of global dust storms in the multi-year Viking Lander pressure records (James et al., 1992). Now the InSight pressure measurements suggest that the Northern seasonal polar cap was slightly and unusually more massive during the winter and spring of MY 34 (after $L_s = 300^\circ$) following an unusual global dust storm that occurred throughout the preceding autumn, well before the observed effect on the seasonal ice cap, in accordance with the observations from MSL (de la Torre Juárez et al., 2019). Based on the discussion above, we can speculate that the most likely reason for this small excess of mass could be due to a slight increase of the ice albedo, resulting from more atmospheric condensation during fall. An alternative explanation could invoke the fact that the post-storm winter atmosphere in the polar night could be slightly depleted in airborne dust and/or ice clouds compared to regular years, reducing the fraction of CO₂ ice clouds and snowfall and therefore increasing the polar night thermal infrared cooling to space, and thus the net condensation rate.

In theory, these hypotheses could be tested using climate simulations performed with a GCM. The current version of the LMD GCM can account for the effect of dust on the atmospheric dynamics and radiative cooling as well as their consequence on the atmospheric CO₂ condensation and its effect on the polar night emissivity (Forget et al., 1998). However, because of the lack of dust observations in the polar night, the dust climatology available to simulate MY 34 (Montabone et al., 2020) in the polar regions is probably not adequate to represent well what happened (either during or after the dust storm). Furthermore, the GCM does not include any feedback on the CO₂ ice deposit albedo, which cannot be affected by dust storm (neither the albedo increase due to the additional atmospheric condensation or decrease by the additional dust contamination). Nevertheless, we performed GCM simulations using the MY 34 and MY 35 dust scenarios (Montabone et al., 2015), looking for other differences that could result from the MY 34 global dust storm. The simulated CO₂ mass cycles in the two years were found to be almost identical (not shown), confirming that processes that are well represented in the GCM (e.g., atmospheric dynamic and heat transport, non-condensable gas enrichment) are probably not involved in the inter-annual seasonal cap variations observed by InSight.

6. Conclusions

In this study, we track long-term pressure changes on Mars by comparing for the first time the InSight and Viking pressure data. We extend this comparison to other pressure data that have been at the surface of Mars over the last 40 years. The main conclusions of this investigation are:

1. InSight pressure measurements have an unexpected thermal sensitivity to sensor temperature, which dramatically impacts the recorded annual pressure and makes its evolution inconsistent over the two years of the mission.
2. A polynomial correction in the sensor temperature is proposed, using a ratio of MSL pressure data to account for the interannual variability of the seasonal pressure cycle, observed by MSL between the beginning of MY 34 and 35. For positive sensor temperature, the correction removes between 1 and 5 Pa to the raw pressure measurements, while it can add between 1 and 15 Pa for negative sensor temperature.
3. InSight data, once recalibrated, have an uncertainty of 1.7–2.3 Pa at $1 - \sigma$ compared to the initial uncertainty of 1.5 Pa at $1 - \sigma$. The correction does not lead to a major uncertainty compromising the detection of secular pressure changes compared to the Viking data, or of interannual changes.
4. The comparison between MSL and InSight pressure during MY 34 and 35 reinforces the credibility of our correction. This comparison also highlights a pressure deficit at the MSL site at $L_s \sim 270^\circ$. This deficit could be induced by a change in the scale height due to a significant amount of dust within Gale Crater, creating a hot atmospheric layer in the local near-surface atmosphere.
5. We design two high-accuracy methods for pressure interpolation, at local and global scales, that correct the effects of local and large-scale atmospheric circulations as well as the martian orography on the seasonal pressure variations. Both methods use a scale height computed with the air temperature at an altitude of 1 km. The influence of atmospheric parameters on this interpolation was quantified at 1% of the absolute pressure at a $3 - \sigma$ level.
6. The Viking 1 and InSight pressure comparison does not show significant secular pressure change, as previously postulated with the Viking and Phoenix comparison. Our results show that the atmospheric mass has not changed by more than ± 8 Pa, knowing that our method has an accuracy of ~ 11 Pa at $3 - \sigma$. This suggests that either the sublimation of the SPRC is much slower than expected, or that the system is actually in equilibrium. In any case, it appears that the mass balance computations that predicted a very large increase in atmospheric mass or the rapid SPRC disappearance are overestimated.
7. Similarly, a first visual comparative analysis of Viking 2 orbiter and MARCI images of the seasonal ice caps does not show significant change in the dynamics of the seasonal ice caps, as observed when comparing the annual variations of the ice caps with pressure data. This comparison requires to be completed with more observations during MY 35 to compute the exact recession curve and compare it with Viking observations.
8. Both of these conclusions are also supported by the comparison between MSL and Viking 1 pressure data. Using the five martian years of MSL pressure records, we cannot establish a secular trend.
9. Phoenix surface pressure data might highlight an increase of the atmospheric mass during MY 29, suggesting a possible erosion of the SPRC after the MY 28 global dust storm. Analysis of the SPRC boundary during MY 27, 28, and 29 would help to study this assumption.
10. The NSPC is more extended during MY 34 compared to MY 35. However, the physical mechanisms that explain this extent are not understood yet. Investigations conducted with the LMD GCM suggest that atmospheric dynamics, heat transport, or non-condensable gas enrichment are not at the origin of this phenomenon.

The Perseverance rover that arrived on Mars on 18 February 2021 at a latitude close to Viking 1 lander will provide a unique new pressure data set to contribute to the study of interannual and secular pressure changes. Cross-analyses between SPRC evolution, dust storms, and atmospheric mass measurements would also help to better understand the evolution of the SPRC and its relative balance.

Appendix A: Impact of MSL Pressure Uncertainties on InSight Pressure Correction

We apply the propagation of uncertainty on the definition of β (Equation 6):

$$\frac{\sigma_\beta}{\beta} = \sqrt{\left(\frac{\sigma_{P_{MSL,Y_1}}}{P_{MSL,Y_1}}\right)^2 + \left(\frac{\sigma_{P_{MSL,Y_2}}}{P_{MSL,Y_2}}\right)^2 - 2 \frac{\text{Cov}(P_{MSL,Y_1}, P_{MSL,Y_2})}{P_{MSL,Y_1} P_{MSL,Y_2}}} \quad (\text{A1})$$

where $\text{Cov}(P_{MSL,Y_1}, P_{MSL,Y_2})$ is the covariance between measurements P_{MSL,Y_1} and P_{MSL,Y_2} .

Let us assume that:

$$P_{MSL,Y_1} = P_{MSL}(t_{Y1}) = P_{atm,true}(t_{Y1}) + \epsilon(t_{Y1}) + \delta(t_{Y1}) \quad (\text{A2})$$

$$P_{MSL,Y_2} = P_{MSL}(t_{Y2}) = P_{atm,true}(t_{Y2}) + \epsilon(t_{Y2}) + \delta(t_{Y2}) \quad (\text{A3})$$

with.

1. $P_{atm,true}(t)$ the true atmospheric pressure that MSL should have recorded without any error
2. $\epsilon(t)$ the error on a measurement due to:
 1. The error on the absolute measurement due to the initial calibration, estimated to be at most 4 Pa at $3 - \sigma$ over the possible pressure range at the MSL landing site (Harri et al., 2014).
 2. The error due to elevation change. During the period considered here, the altitude of the rover changed by nearly 100 m, which could lead to a change in pressure of 8 Pa at $3 - \sigma$.
 3. At least, the estimated error is $\sqrt{8^2 + 4^2} \approx 9$ Pa at $3 - \sigma$
 4. Since 15-day averaged data are used, the uncertainty related to the precision of the measurements is assumed to be negligible.
3. $\delta(t)$, the drift error which theoretically evolves at a rate of 1 Pa/MY at $3 - \sigma$ (Harri et al., 2014).

We model these errors by random variables whose variance is given by the previous values. The last two terms in the expression of P_{MSL} , thus representing the error on the measurement, are random variables of variance. We introduce $\sigma_{P_{MSL}} = \sqrt{\sigma_\epsilon^2 + \sigma_\delta^2}$ by independence of these two terms.

Since the errors are computed over the range of possible values of the MSL measurements and not a precise value, the errors ϵ , δ are independent of $P_{atm,true}$. Moreover, the $P_{atm,true}$ between the two years are completely independent. Using the bilinearity of the covariance, and these independences, we obtain:

$$\text{Cov}(P_{MSL,Y_1}, P_{MSL,Y_2}) = \text{Cov}(\epsilon(t_{Y1}), \epsilon(t_{Y2})) + \text{Cov}(\delta(t_{Y1}), \delta(t_{Y2})) \quad (\text{A4})$$

By definition, for two random variables a , b :

$$\text{Cov}(a, b) = \rho(a, b)\sigma_a\sigma_b \quad (\text{A5})$$

with ρ the correlation coefficient. ϵ has been determined during calibration tests and is assumed to be constant over the mission, so that

$$\text{Cov}(\epsilon(t_{Y1}), \epsilon(t_{Y2})) = \sigma_\epsilon^2 \quad (\text{A6})$$

Assuming that the drift grows at a rate of 1 Pa/MY, we have:

$$\text{Cov}(\delta(t_{Y1}), \delta(t_{Y2})) = \text{Cov}(\delta(t_{Y1}), \delta(t_{Y1}) + 1) = \text{Cov}(\delta(t_{Y1}), \delta(t_{Y1})) \quad (\text{A7})$$

by property of the covariance. We thus have:

$$\text{Cov}(\delta(t_{Y1}), \delta(t_{Y2})) = \sigma_\delta^2 \quad (\text{A8})$$

Hence, Equation A4 becomes:

$$\frac{\sigma_\beta}{\beta} = \sigma_{P_{MSL}} \sqrt{\left(\frac{1}{P_{MSL,Y1}} - \frac{1}{P_{MSL,Y2}}\right)^2} \quad (\text{A9})$$

which gives:

$$\frac{\sigma_\beta}{\beta} \sim 5 \times 10^{-5} \quad (\text{A10})$$

Data Availability Statement

InSight pressure uncorrected data can be retrieved from the PDS (Banfield, 2019), MSL REMS and Viking 1 pressure data can be retrieved from the PDS (Gomez-Elvira, 2019; Tillman, 1989). Phoenix corrected data are given with Taylor et al. (2010). MARCI mosaics can be reconstructed from the images that can be retrieved from the PDS (Malin, 2007). THEMIS images can be retrieved from the PDS (Christensen, 2002). The Mars Climate Database can be retrieved upon request (see <http://www-mars.lmd.jussieu.fr/mars/access.html>). Data files for figures used in this analysis are available in a public repository, see Lange et al. (2022).

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References

- Ball, E. R., Mitchell, D. M., Seviour, W. J. M., Thomson, S. I., & Vallis, G. K. (2021). The roles of latent heating and dust in the structure and variability of the northern martian polar vortex. *The Planetary Science Journal*, 2(5), 203. <https://doi.org/10.3847/psj/ac1ba2>
- Banerdt, W. B., Smrekar, S. E., Banfield, D., Giardini, D., Golombek, M., Johnson, C. L., et al. (2020). Initial results from the InSight mission on Mars. *Nature Geoscience*, 13(3), 183–189. <https://doi.org/10.1038/s41561-020-0544-y>
- Banfield, D. (2019). InSight APSS PS data product bundle. *NASA Planetary Data System*. <https://doi.org/10.17189/1518939>
- Banfield, D., Rodriguez-Manfredi, J. A., Russell, C. T., Rowe, K. M., Leneman, D., Lai, H. R., et al. (2019). InSight auxiliary payload sensor suite (APSS). *Space Science Reviews*, 215(1), 4. <https://doi.org/10.1007/s11214-018-0570-x>
- Banfield, D., Spiga, A., Newman, C., Forget, F., Lemmon, M., Lorenz, R., et al. (2020). The atmosphere of Mars as observed by InSight. *Nature Geoscience*, 13(3), 190–198. <https://doi.org/10.1038/s41561-020-0534-0>
- Barnes, J. R. (1980). Time spectral analysis of midlatitude disturbances in the martian atmosphere. *Journal of the Atmospheric Sciences*, 37(9), 2002–2015. [https://doi.org/10.1175/1520-0469\(1980\)037<2002:tsaomd>2.0.co;2](https://doi.org/10.1175/1520-0469(1980)037<2002:tsaomd>2.0.co;2)
- Becerra, P., Byrne, S., & Brown, A. J. (2015). Transient bright “halos” on the south polar residual cap of Mars: Implications for mass-balance. *Icarus*, 251, 211–225. <https://doi.org/10.1016/j.icarus.2014.04.050>
- Benson, J. L., & James, P. B. (2005). Yearly comparisons of the martian polar caps: 1999–2003 Mars orbiter camera observations. *Icarus*, 174(2), 513–523. (Mars Polar Science III). <https://doi.org/10.1016/j.icarus.2004.08.025>
- Blackburn, D. G., Bryson, K. L., Chevrier, V. F., Roe, L. A., & White, K. F. (2010). Sublimation kinetics of CO₂ ice on Mars. *Planetary and Space Science*, 58(5), 780–791. <https://doi.org/10.1016/j.pss.2009.12.004>
- Bonev, B. P. (2002). Regression of the mountains of Mitchel polar ice after the onset of a global dust storm on Mars. *Geophysical Research Letters*, 29(21), 2017. <https://doi.org/10.1029/2002gl015458>
- Bonev, B. P., Hansen, G. B., Glenar, D. A., James, P. B., & Bjorkman, J. E. (2008). Albedo models for the residual south polar cap on Mars: Implications for the stability of the cap under near-perihelion global dust storm conditions. *Planetary and Space Science*, 56(2), 181–193. (Mars Polar Processes: Atmosphere-Surface Interactions). <https://doi.org/10.1016/j.pss.2007.08.003>
- Bonev, B. P., James, P. B., Bjorkman, J. E., Hansen, G. B., & Wolff, M. J. (2003). Effects of atmospheric and surface dust on the sublimation rates of CO₂ on Mars. In S. Clifford, P. Doran, D. Fisher, & C. Herd (Eds.), *Third international conference on Mars polar science and exploration* (p. 8052).
- Brown, A. J., Calvin, W. M., McGuire, P. C., & Murchie, S. L. (2010). Compact Reconnaissance Imaging Spectrometer for Mars (CRISM) south polar mapping: First Mars year of observations. *Journal of Geophysical Research*, 115(E2), E00D13. <https://doi.org/10.1029/2009JE003333>
- Brown, A. J., Calvin, W. M., & Murchie, S. L. (2012). Compact Reconnaissance Imaging Spectrometer for Mars (CRISM) north polar springtime recession mapping: First 3 Mars years of observations. *Journal of Geophysical Research*, 117(E12). <https://doi.org/10.1029/2012JE004113>
- Buhler, P. B., Ingersoll, A. P., Piqueux, S., Ehlmann, B. L., & Hayne, P. O. (2020). Coevolution of Mars's atmosphere and massive south polar CO₂ ice deposit. *Nature Astronomy*, 4(4), 364–371. <https://doi.org/10.1038/s41550-019-0976-8>
- Byrne, S., Hayne, P. O., Becerra, P., & HiRISE Team. (2015). Evolution and stability of the residual CO₂ ice cap. In *Lunar and planetary science conference* (p. 1657).
- Byrne, S., Zuber, M. T., & Neumann, G. A. (2008). Interannual and seasonal behavior of Martian residual ice-cap albedo. *Planetary and Space Science*, 56(2), 194–211. <https://doi.org/10.1016/j.pss.2006.03.018>
- Calvin, W., Cantor, B., & James, P. (2017). Interannual and seasonal changes in the south seasonal polar cap of Mars: Observations from MY 28–31 using MARCI. *Icarus*, 292, 144–153. <https://doi.org/10.1016/j.icarus.2017.01.010>
- Calvin, W., James, P., Cantor, B., & Dixon, E. (2015). Interannual and seasonal changes in the north polar ice deposits of Mars: Observations from MY 29–31 using MARCI. *Icarus*, 251, 181–190. (Dynamic Mars). <https://doi.org/10.1016/j.icarus.2014.08.026>
- Chatain, A., Spiga, A., Banfield, D., Forget, F., & Murdoch, N. (2021). Seasonal variability of the daytime and nighttime atmospheric turbulence experienced by InSight on Mars. *Geophysical Research Letters*, 48(22), e2021GL095453. e2021GL095453 2021GL095453. <https://doi.org/10.1029/2021GL095453>
- Christensen, P. (2002). Odyssey THEMIS-VIS geo v2.0. *NASA Planetary Data System*. <https://doi.org/10.17189/1520386>
- Christensen, P. R., Jakosky, B. M., Kieffer, H. H., Malin, M. C., Harry, Y., McSweeney, J., et al. (2004). The thermal emission imaging system (THEMIS) for the Mars 2001 odyssey mission. *Space Science Reviews*, 110(1/2), 85–130. <https://doi.org/10.1023/b:spac.0000021008.16305.94>
- Cornwall, C., & Titus, T. N. (2009). Spatial and temporal distributions of martian north polar cold spots before, during, and after the global dust storm of 2001. *Journal of Geophysical Research*, 114(E2), E02003. <https://doi.org/10.1029/2008je003243>
- de la Torre Juarez, M., Piqueux, S., Kass, D. M., Newman, C., & Guzewich, S. D. (2019). Pressure deficit in Gale crater and a larger northern polar cap after the Mars year 34 global dust storm. *AGU fall meeting abstracts*, 2019, P51C–02.
- Fonseca, R. M., Zorzano-Mier, M.-P., & Martín-Torres, J. (2018). Planetary boundary layer and circulation dynamics at Gale crater, Mars. *Icarus*, 302, 537–559. <https://doi.org/10.1016/j.icarus.2017.11.036>
- Forget, F., Hourdin, F., Fournier, R., Hourdin, C., Talagrand, O., Collins, M., et al. (1999). Improved general circulation models of the martian atmosphere from the surface to above 80 km. *Journal of Geophysical Research*, 104(E10), 24155–24175. <https://doi.org/10.1029/1999JE001025>
- Forget, F., Hourdin, F., & Talagrand, O. (1998). CO₂ snowfall on Mars: Simulation with a general circulation model. *Icarus*, 131(2), 302–316. <https://doi.org/10.1006/icar.1997.5874>
- Forget, F., Millour, E., Montabone, L., & Lefevre, F. (2008). Non condensable gas enrichment and depletion in the martian polar regions. *Third international workshop on the Mars atmosphere: Modeling and observations*, 1447, 9106.

- Forget, F., & Pollack, J. B. (1996). Thermal infrared observations of the condensing martian polar caps: CO₂ ice temperatures and radiative budget. *Journal of Geophysical Research*, 101(E7), 16865–16879. <https://doi.org/10.1029/96je01077>
- Forget, F., Spiga, A., Dolla, B., Vinatier, S., Melchiorri, R., Drossart, P., et al. (2007). Remote sensing of surface pressure on Mars with the Mars express/OMEGA spectrometer: 1. Retrieval method. *Journal of Geophysical Research*, 112(E8). <https://doi.org/10.1029/2006JE002871>
- Giuranna, M., Formisano, V., Grassi, D., & Maturilli, A. (2007). Tracking the edge of the south seasonal polar cap of Mars. *Planetary and Space Science*, 55(10), 1319–1327. <https://doi.org/10.1016/j.pss.2006.12.005>
- Golombek, M., Williams, N., Warner, N. H., Parker, T., Williams, M. G., Daubar, I., et al. (2020). Location and setting of the Mars InSight lander, instruments, and landing site. *Earth and Space Science*, 7(10), e2020EA001248. <https://doi.org/10.1029/2020EA001248>
- Gomez-Elvira, J. (2019). Mars Science Laboratory rover environmental monitoring station rdr data v1.0, modrdr v1.0. *NASA Planetary Data System*. <https://doi.org/10.17189/1523033>
- Gómez-Elvira, J., Armiens, C., Carrasco, I., Genzer, M., Gómez, F., Haberle, R., et al. (2014). Curiosity's rover environmental monitoring station: Overview of the first 100 sols. *Journal of Geophysical Research: Planets*, 119(7), 1680–1688. <https://doi.org/10.1002/2013JE004576>
- Guzewich, S. D., Newman, C. E., Smith, M. D., Moores, J. E., Smith, C. L., Moore, C., et al. (2017). The vertical dust profile over Gale crater, Mars. *Journal of Geophysical Research: Planets*, 122(12), 2779–2792. <https://doi.org/10.1002/2017je005420>
- Guzewich, S. D., Toigo, A., & Waugh, D. (2016). The effect of dust on the martian polar vortices. *Icarus*, 278, 100–118. <https://doi.org/10.1016/j.icarus.2016.06.009>
- Haberle, R. M., Gómez-Elvira, J., de la Torre Juárez, M., Harri, A.-M., Hollingsworth, J. L., Kahanpää, H., et al. (2014). Preliminary interpretation of the REMS pressure data from the first 100 sols of the MSL mission. *Journal of Geophysical Research: Planets*, 119(3), 440–453. <https://doi.org/10.1002/2013JE004488>
- Haberle, R. M., & Kahre, M. A. (2010). Detecting secular climate change on Mars. *International Journal of Mars Science and Exploration*, 4, 68–75. <https://doi.org/10.1555/mars.2010.0003>
- Harri, A. M., Genzer, M., Kemppinen, O., Kahanpää, H., Gomez-Elvira, J., Rodriguez-Manfredi, J. A., et al. (2014). Pressure observations by the Curiosity rover: Initial results. *Journal of Geophysical Research*, 119(1), 82–92. <https://doi.org/10.1002/2013JE004423>
- Hess, S. L., Henry, R. M., Leovy, C. B., Ryan, J. A., Tillman, J. E., Chamberlain, T. E., et al. (1976). Preliminary meteorological results on Mars from the Viking 1 lander. *Science*, 193(4255), 788–791. <https://doi.org/10.1126/science.193.4255.788>
- Hess, S. L., Ryan, J. A., Tillman, J. E., Henry, R. M., & Leovy, C. B. (1980). The annual cycle of pressure on Mars measured by Viking landers 1 and 2. *Geophysical Research Letters*, 7(3), 197–200. <https://doi.org/10.1029/gl007i003p00197>
- Holstein-Rathlou, C., Gunnlaugsson, H., Cantor, B., Ellehöj, M., Lange, C., Lemmon, M., et al. (2010). On dust storms observed at the Phoenix landing site. *41st Annual Lunar and Planetary Science Conference*, 1533, 1837. (Conference date: 01-03-2010 through 05-03-2010).
- Hourdin, F., Forget, F., & Talagrand, O. (1995). The sensitivity of the martian surface pressure and atmospheric mass budget to various parameters: A comparison between numerical simulations and viking observations. *Journal of Geophysical Research*, 100(E3), 5501–5523. <https://doi.org/10.1029/94JE03079>
- Hourdin, F., Van, P. L., Forget, F., & Talagrand, O. (1993). Meteorological variability and the annual surface pressure cycle on Mars. *Journal of the Atmospheric Sciences*, 50(21), 3625–3640. [https://doi.org/10.1175/1520-0469\(1993\)050<3625:mvtas>2.0.co;2](https://doi.org/10.1175/1520-0469(1993)050<3625:mvtas>2.0.co;2)
- Jakosky, B. M., & Haberle, R. M. (1990). Year-to-year instability of the Mars south polar cap. *Journal of Geophysical Research*, 95(B2), 1359. <https://doi.org/10.1029/jb095ib02p01359>
- James, P., Cantor, B., Malin, M., Edgett, K., Carr, M., Danielson, G., & Veverka, J. (2000). The 1997 spring regression of the martian South Polar cap: Mars orbiter camera observations. *Icarus*, 144(2), 410–418. <https://doi.org/10.1006/icar.1999.6289>
- James, P., Thomas, P., & Malin, M. (2010). Variability of the south polar cap of Mars in Mars Years 28 and 29. *Icarus*, 208(1), 82–85. <https://doi.org/10.1016/j.icarus.2010.02.007>
- James, P. B. (1979). Recession of martian north polar cap: 1977–1978 viking observations. *Journal of Geophysical Research*, 84(B14), 8332. <https://doi.org/10.1029/jb084ib14p08332>
- James, P. B. (1982). Recession of martian north polar cap: 1979–1980 viking observations. *Icarus*, 52(3), 565–569. [https://doi.org/10.1016/0019-1035\(82\)90016-1](https://doi.org/10.1016/0019-1035(82)90016-1)
- James, P. B., Briggs, G., Barnes, J., & Spruck, A. (1979). Seasonal recession of Mars' south polar cap as seen by Viking. *Journal of Geophysical Research*, 84(B6), 2889. <https://doi.org/10.1029/jb084ib06p02889>
- James, P. B., & Cantor, B. A. (2001). *Martian North Polar Cap Recession: 2000 Mars Orbiter Camera Observations*, 154(1), 131–144. <https://doi.org/10.1006/icar.2001.6653>
- James, P. B., Cantor, B. A., & Davis, S. (2001). Mars Orbiter Camera observations of the martian south polar cap in 1999–2000. *Journal of Geophysical Research*, 106(E10), 23635–23652. <https://doi.org/10.1029/2000JE001313>
- James, P. B., Clancy, R., Lee, S. W., Martin, L. J., & Bell, J. (1996). Seasonal recession of martian South Polar cap: 1992 HST observations. *Icarus*, 123(1), 87–100. <https://doi.org/10.1006/icar.1996.0143>
- James, P. B., Kieffer, H. H., & Paige, D. A. (1992). The seasonal cycle of carbon dioxide on Mars. In M. George (Ed.), *Mars*, 123, 934–968.
- James, P. B., & Lumme, K. (1982). Martian South Polar cap boundary: 1971 and 1973 data. *Icarus*, 50(2–3), 368–380. [https://doi.org/10.1016/0019-1035\(82\)90130-0](https://doi.org/10.1016/0019-1035(82)90130-0)
- Kahanpää, H., Newman, C., Moores, J., Zorzano, M.-P., Martín-Torres, J., Navarro, S., et al. (2016). Convective vortices and dust devils at the MSL landing site: Annual variability. *Journal of Geophysical Research: Planets*, 121(8), 1514–1549. <https://doi.org/10.1002/2016JE005027>
- Kahanpää, H., Polkko, J., & Daly, M. (2021). Accuracy of the Phoenix and viking atmospheric pressure measurements: Impact on detecting the climate change on Mars. *European Planetary Science Congress*. EPSC2021-63. <https://doi.org/10.5194/epsc2021-63>
- Kahre, M. A., & Haberle, R. M. (2010). Mars CO₂ cycle: Effects of airborne dust and polar cap ice emissivity. *Icarus*, 207(2), 648–653. <https://doi.org/10.1016/j.icarus.2009.12.016>
- Kahre, M. A., Murphy, J. R., Newman, C. E., Wilson, R. J., Cantor, B. A., Lemmon, M. T., & Wolff, M. J. (2017). The Mars dust cycle. In R. M. Haberle, R. T. Clancy, F. Forget, M. D. Smith, & R. W. Zurek (Eds.), *The atmosphere and climate of Mars* (pp. 295–337). Cambridge University Press. <https://doi.org/10.1017/9781139060172.010>
- Kieffer, H. H. (1979). Mars South Polar spring and summer temperatures: A residual CO₂ frost. *Journal of Geophysical Research*, 84(B14), 8263–8288. <https://doi.org/10.1029/jb084ib14p08263>
- Kieffer, H. H., Neugebauer, G., Munch, G., Chase, S. C., & Miner, E. (1972). Infrared thermal mapping experiment: The Viking Mars orbiter. *Icarus*, 16(1), 47–56. [https://doi.org/10.1016/0019-1035\(72\)90136-4](https://doi.org/10.1016/0019-1035(72)90136-4)
- Kieffer, H. H., Titus, T. N., Mullins, K. F., & Christensen, P. R. (2000). Mars south polar spring and summer behavior observed by TES. *Journal of Geophysical Research*, 105(E4), 9653–9699. <https://doi.org/10.1029/1999je001136>
- Lange, L., Forget, F., Banfield, D., Wolff, M., Spiga, A., Millour, E., et al. (2022). InSight pressure data recalibration, and its application to the study of long-term pressure changes on Mars. *Harvard Dataverse*. <https://doi.org/10.7910/DVN/RUWICY>

- Langevin, Y., Bibring, J.-P., Montmessin, F., Forget, F., Vincendon, M., Douté, S., et al. (2007). Observations of the south seasonal cap of Mars during recession in 2004–2006 by the OMEGA visible/near-infrared imaging spectrometer on board Mars Express. *Journal of Geophysical Research*, 112(E8), E08S12. <https://doi.org/10.1029/2006JE002841>
- Leighton, R. B., & Murray, B. C. (1966). Behavior of carbon dioxide and other volatiles on Mars. *Science*, 153(3732), 136–144. <https://doi.org/10.1126/science.153.3732.136>
- Lemmon, M. T., Guzewich, S. D., McConnochie, T., de Vicente-Retortillo, A., Martínez, G., Smith, M. D., et al. (2019). Large dust aerosol sizes seen during the 2018 martian global dust event by the Curiosity rover. *Geophysical Research Letters*, 46(16), 9448–9456. <https://doi.org/10.1029/2019GL084407>
- Malin, M. (2007). MRO Mars COLOR IMAGER experiment data record level 0 v1.0. *NASA Planetary Data System*. <https://doi.org/10.17189/1520403>
- Malin, M. C., Bell, J. F., III, Calvin, W., Clancy, R. T., Haberle, R. M., James, P. B., et al. (2001). Mars color imager (MARCI) on the Mars climate orbiter. *Journal of Geophysical Research: Planets*, 106(E8), 17651–17672. <https://doi.org/10.1029/1999je001145>
- Malin, M. C., Caplinger, M. A., & Davis, S. D. (2001). Observational evidence for an active surface reservoir of solid carbon dioxide on Mars. *Science*, 294(5549), 2146–2148. <https://doi.org/10.1126/science.1066416>
- Martínez, G., Fischer, E., Rennó, N., Sebastián, E., Kemppinen, O., Bridges, N., et al. (2016). Likely frost events at Gale crater: Analysis from MSL/REMS measurements. *Icarus*, 280, 93–102. (MicroMars to MegaMars). <https://doi.org/10.1016/j.icarus.2015.12.004>
- Martínez, G. M., Newman, C. N., De Vicente-Retortillo, A., Fischer, E., Renno, N. O., Richardson, M. I., et al. (2017). The modern near-surface martian climate: A review of in-situ meteorological data from viking to curiosity. *Space Science Reviews*, 212(1–2), 295–338. <https://doi.org/10.1007/s11214-017-0360-x>
- McCleese, D. J., Schofield, J. T., Taylor, F. W., Calcutt, S. B., Foote, M. C., Kass, D. M., et al. (2007). Mars Climate Sounder: An investigation of thermal and water vapor structure, dust and condensate distributions in the atmosphere, and energy balance of the polar regions. *Journal of Geophysical Research*, 112(E5), E05S06. <https://doi.org/10.1029/2006JE002790>
- Millour, E., Forget, F., Spiga, A., Vals, M., Zakharov, V., Montabone, L., et al. (2018). The Mars climate Database (version 5.3). In *From Mars express to ExoMars* (p. 68).
- Mitchell, M. (1977). *Evaluation of Viking lander barometric pressure sensor*. (No. NASA-TM-X-74020).
- Montabone, L., Forget, F., Millour, E., Wilson, R., Lewis, S., Cantor, B., et al. (2015). Eight-year climatology of dust optical depth on Mars. *Icarus*, 251, 65–95. (Dynamic Mars). <https://doi.org/10.1016/j.icarus.2014.12.034>
- Montabone, L., Spiga, A., Kass, D. M., Kleinböhl, A., Forget, F., & Millour, E. (2020). Martian year 34 column dust climatology from Mars Climate Sounder observations: Reconstructed maps and model simulations. *Journal of Geophysical Research: Planets*, 125(8), e2019JE006111. <https://doi.org/10.1029/2019JE006111>
- Moore, C. A., Moores, J. E., Newman, C. E., Lemmon, M. T., Guzewich, S. D., & Battalio, M. (2019). Vertical and horizontal heterogeneity of atmospheric dust loading in northern Gale crater, Mars. *Icarus*, 329, 197–206. <https://doi.org/10.1016/j.icarus.2019.03.041>
- Moores, J. E., Lemmon, M. T., Kahanpää, H., Rafkin, S. C., Francis, R., Pla-Garcia, J., et al. (2015). Observational evidence of a suppressed planetary boundary layer in northern Gale crater, Mars as seen by the Navcam instrument onboard the Mars Science Laboratory rover. *Icarus*, 249, 129–142. (Special Issue: First Year of MSL). <https://doi.org/10.1016/j.icarus.2014.09.020>
- Morris, E. C., & Jones, K. L. (1980). Viking 1 lander on the surface of Mars: Revised location. *Icarus*, 44(1), 217–222. [https://doi.org/10.1016/0019-1035\(80\)90067-6](https://doi.org/10.1016/0019-1035(80)90067-6)
- Newman, C. E., Kahanpää, H., Richardson, M. I., Martínez, G. M., Vicente-Retortillo, A., & Lemmon, M. T. (2019). MarsWRF convective vortex and dust devil predictions for Gale crater over 3 Mars years and comparison with MSL-REMS observations. *Journal of Geophysical Research: Planets*, 124(12), 3442–3468. <https://doi.org/10.1029/2019JE006082>
- Ordóñez-Exteberria, I., Hueso, R., & Sánchez-Lavega, A. (2018). A systematic search of sudden pressure drops on Gale crater during two Martian years derived from MSL/REMS data. *Icarus*, 299, 308–330. <https://doi.org/10.1016/j.icarus.2017.07.032>
- Ordóñez-Exteberria, I., Hueso, R., Sánchez-Lavega, A., Millour, E., & Forget, F. (2019). Meteorological pressure at Gale crater from a comparison of REMS/MSL data and MCD modelling: Effect of dust storms. *Icarus*, 317, 591–609. <https://doi.org/10.1016/j.icarus.2018.09.003>
- Piqueux, S., & Christensen, P. R. (2008). Deposition of CO₂ and erosion of the martian south perennial cap between 1972 and 2004: Implications for current climate change. *Journal of Geophysical Research*, 113(E2), E02006. <https://doi.org/10.1029/2007JE002969>
- Piqueux, S., Kleinböhl, A., Hayne, P. O., Kass, D. M., Heavens, N., McCleese, D. J., et al. (2020). Atmospheric CO₂ depletion at the surface in the polar regions of Mars. *Seventh international conference on Mars polar science and exploration*, 2099, 6016.
- Piqueux, S., Kleinböhl, A., Hayne, P. O., Kass, D. M., Schofield, J. T., & McCleese, D. J. (2015). Variability of the martian seasonal CO₂ cap extent over eight Mars years. *Icarus*, 251, 164–180. (Dynamic Mars). <https://doi.org/10.1016/j.icarus.2014.10.045>
- Pla-Garcia, J., Rafkin, S. C., Kahre, M., Gomez-Elvira, J., Hamilton, V. E., Navarro, S., et al. (2016). The meteorology of Gale crater as determined from rover environmental monitoring station observations and numerical modeling. part i: Comparison of model simulations with observations. *Icarus*, 280, 103–113. (MicroMars to MegaMars). <https://doi.org/10.1016/j.icarus.2016.03.013>
- Pollack, J. B., Haberle, R. M., Schaeffer, J., & Lee, H. (1990). Simulations of the general circulation of the martian atmosphere: 1. Polar processes. *Journal of Geophysical Research*, 95(B2), 1447. <https://doi.org/10.1029/jb095ib02p01447>
- Press, W. H., Teukolsky, S. A., Vetterling, W. T., & Flannery, B. P. (1993). *Numerical recipes in FORTRAN; the art of scientific computing* (2nd ed.). Cambridge University Press.
- Rafkin, S. C., Pla-Garcia, J., Kahre, M., Gomez-Elvira, J., Hamilton, V. E., Marín, M., et al. (2016). The meteorology of Gale Crater as determined from rover environmental monitoring station observations and numerical modeling. Part ii: Interpretation. *Icarus*, 280, 114–138. (MicroMars to MegaMars). <https://doi.org/10.1016/j.icarus.2016.01.031>
- Rangarajan, V. G., & Ghosh, M. (2020). Seasonal thermal inertia variations at Gale crater: Role of active surface deposition phenomena. *Icarus*, 337, 113499. <https://doi.org/10.1016/j.icarus.2019.113499>
- Richardson, M. I., & Newman, C. E. (2018). On the relationship between surface pressure, terrain elevation, and air temperature. Part i: The large diurnal surface pressure range at Gale crater, Mars and its origin due to lateral hydrostatic adjustment. *Planetary and Space Science*, 164, 132–157. <https://doi.org/10.1016/j.pss.2018.07.003>
- Ryan, J. A., & Sharman, R. D. (1981). Two major dust storms, one Mars year apart: Comparison from Viking data. *Journal of Geophysical Research*, 86(C4), 3247–3254. <https://doi.org/10.1029/JC086iC04p03247>
- Seiff, A. (1976). The Viking atmosphere structure experiment - Techniques, instruments, and expected accuracies. *Space Science Instrumentation*, 2, 381–423.
- Seiff, A., & Kirk, D. B. (1977). Structure of the atmosphere of Mars in summer at mid-latitudes. *Journal of Geophysical Research*, 82(28), 4364–4378. <https://doi.org/10.1029/JS082i028p04364>

- Spiga, A., Banfield, D., Teanby, N. A., Forget, F., Lucas, A., Kenda, B., et al. (2018). Atmospheric science with InSight. *Space Science Reviews*, 214(7), 109. <https://doi.org/10.1007/s11214-018-0543-0>
- Spiga, A., & Forget, F. (2009). A new model to simulate the martian mesoscale and microscale atmospheric circulation: Validation and first results. *Journal of Geophysical Research*, 114(E2), E02009. <https://doi.org/10.1029/2008JE003242>
- Spiga, A., Forget, F., Dolla, B., Vinatier, S., Melchiorri, R., Drossart, P., et al. (2007). Remote sensing of surface pressure on Mars with the Mars express/OMEGA spectrometer: 2. Meteorological maps. *Journal of Geophysical Research*, 112(E8). <https://doi.org/10.1029/2006JE002870>
- Spiga, A., Murdoch, N., Lorenz, R., Forget, F., Newman, C., Rodriguez, S., et al. (2021). A study of daytime convective vortices and turbulence in the martian planetary boundary layer based on half-a-year of insight atmospheric measurements and large-eddy simulations. *Journal of Geophysical Research: Planets*, 126(1), e2020JE006511. <https://doi.org/10.1029/2020JE006511>
- Steakley, K., & Murphy, J. (2016). A year of convective vortex activity at Gale crater. *Icarus*, 278, 180–193. <https://doi.org/10.1016/j.icarus.2016.06.010>
- Streeter, P. M., Lewis, S. R., Patel, M. R., Holmes, J. A., Fedorova, A. A., Kass, D. M., & Kleinböhl, A. (2021). Asymmetric impacts on Mars' polar vortices from an equinoctial global dust storm. *Journal of Geophysical Research: Planets*, 126(5), e2020JE006774. <https://doi.org/10.1029/2020JE006774>
- Taylor, P. A., Kahanpää, H., Weng, W., Akingunola, A., Cook, C., Daly, M., et al. (2010). On pressure measurement and seasonal pressure variations during the Phoenix mission. *Journal of Geophysical Research*, 115(E3), E00E15. <https://doi.org/10.1029/2009JE003422>
- Thomas, P., Calvin, W., Cantor, B., Haberle, R., James, P., & Lee, S. (2016). Mass balance of Mars' residual south polar cap from CTX images and other data. *Icarus*, 268, 118–130. <https://doi.org/10.1016/j.icarus.2015.12.038>
- Thomas, P., Calvin, W., Gierasch, P., Haberle, R., James, P., & Sholes, S. (2013). Time scales of erosion and deposition recorded in the residual south polar cap of Mars. *Icarus*, 225(2), 923–932. (Mars Polar Science V). <https://doi.org/10.1016/j.icarus.2012.08.038>
- Thomas, P., James, P., Calvin, W., Haberle, R., & Malin, M. (2009). Residual south polar cap of Mars: Stratigraphy, history, and implications of recent changes. *Icarus*, 203(2), 352–375. <https://doi.org/10.1016/j.icarus.2009.05.014>
- Tillman, J. (1989). VL1/vL2 Mars meteorology resampled daily average pressure V1.0. *NASA Planetary Data System*. <https://doi.org/10.17189/ME53-8Z53>
- Tillman, J. E. (1988). Mars global atmospheric oscillations: Annually synchronized, transient normal-mode oscillations and the triggering of global dust storms. *Journal of Geophysical Research*, 93(D8), 9433–9451. <https://doi.org/10.1029/JD093iD08p09433>
- Tillman, J. E., Johnson, N. C., Guttorp, P., & Percival, D. B. (1993). The martian annual atmospheric pressure cycle: Years without great dust storms. *Journal of Geophysical Research*, 98(E6), 10963–10971. <https://doi.org/10.1029/93JE01084>
- Titus, T. N., Byrne, S., Colaprete, A., Forget, F., Michaels, T. I., & Prettyman, T. H. (2017). The CO₂ cycle. In R. M. Haberle, R. T. Clancy, F. Forget, M. D. Smith, & R. W. Zurek (Eds.), *The atmosphere and climate of Mars* (pp. 374–404). Cambridge University Press. <https://doi.org/10.1017/9781139060172.012>
- Vicente-Retortillo, Á., Martínez, G. M., Renno, N., Newman, C. E., Ordonez-Etxeberria, I., Lemmon, M. T., et al. (2018). Seasonal deposition and lifting of dust on Mars as observed by the Curiosity rover. *Scientific Reports*, 8(1), 17576. <https://doi.org/10.1038/s41598-018-35946-8>
- Viúdez-Moreiras, D., Newman, C. E., de la Torre, M., Martínez, G., Guzewich, S., Lemmon, M., et al. (2019). Effects of the MY34/2018 global dust storm as measured by MSL REMS in Gale crater. *Journal of Geophysical Research: Planets*, 124(7), 1899–1912. <https://doi.org/10.1029/2019JE005985>
- Viúdez-Moreiras, D., Newman, C. E., Forget, F., Lemmon, M., Banfield, D., Spiga, A., et al. (2020). Effects of a large dust storm in the near-surface atmosphere as measured by InSight in Elysium Planitia, Mars. comparison with contemporaneous measurements by Mars Science Laboratory. *Journal of Geophysical Research: Planets*, 125(9). <https://doi.org/10.1029/2020je006493>
- Withers, P. (2012). Empirical estimates of martian surface pressure in support of the landing of Mars. *Science Laboratory*, 170(1–4), 837–860. <https://doi.org/10.1007/s11214-012-9876-2>
- Zhao, Y., Zhong, L., Yuan, R., Zhao, C., Li, R., Wang, Y., et al. (2021). Simulation of martian dust effects on polar CO₂ ice caps and atmospheric circulation using the marswrf model. *Journal of Geophysical Research: Planets*, 126(12), e2021JE006937. (e2021JE006937 2021JE006937). <https://doi.org/10.1029/2021JE006937>