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The Determination of the Rotational State and Interior Structure of Venus with VERITAS

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Abstract

Understanding the processes that led Venus to its current state and will drive its future evolution is a major objective of the next generation of orbiters. In this work we analyze the retrieval of the spin vector, the tidal response, and the moment of inertia of Venus with VERITAS, a NASA Discovery-class mission. By simulating a systematic joint analysis of Doppler tracking data and tie points provided by the onboard synthetic aperture radar, we show that VERITAS will provide accuracies (3σ) in the estimates of the tidal Love number k_2 to 4.6×10^{-4} , its tidal phase lag to 0.05° , and the moment of inertia factor to 9.8×10^{-4} (0.3% of the expected value). Applying these results to recent models of the Venus interior, we show that VERITAS will provide much-improved constraints on the interior structure of the planet.

Unified Astronomy Thesaurus concepts: Venus (1763); Solar system terrestrial planets (797); Orbit determination (1175); Planetary interior (1248); Planetary structure (1256); Remote sensing (2191); Planetary science (1255)

1. Introduction

The most comprehensive mapping of Venus was done by the Magellan mission in the early 1990s (Saunders et al. 1992). Magellan employed a combination of data from Doppler tracking and S-band Synthetic Aperture Radar (SAR), an altimeter and radiometer to make nearly global observations of the surface of Venus (Ford & Pettengill 1992). Magellan observations led to the most accurate in situ estimate of the planet's spin axis orientation, sidereal rotation period (Davies et al. 1992; see also Campbell et al. 2019 for a summary of several other observation campaigns), gravity field, and tidal response (Konopliv et al. 1999; Konopliv & Yoder 1996). The Magellan estimates, however, proved not sufficiently precise to constrain the structure of the mantle and core. As shown in Dumoulin et al. (2017), current estimates of the tidal response do not distinguish between a liquid and solid core, and the absence of a measurement of the tidal phase lag prevents us from inferring the viscous response of the interior. Until recently, models of Venus's interior relied solely on scaling Earth's interior structure to Venus's radius (e.g., Yoder 1995; Aitta 2012). A recent direct (ground-based) measurement of the moment of inertia factor ($\text{MOIF} = C/MR^2$, where C is the polar moment of inertia and M and R are the planetary mass and radius, respectively) yields 7% fractional uncertainty and provides weak constraints on the internal density profile and core size (Margot et al. 2021). Improved measurements are needed to quantify the interior structure of Venus with precision.

The Venus Emissivity, Radio science, INSAR, Topography And Spectroscopy (VERITAS) mission (Freeman & Smrekar 2015; Smrekar & the VERITAS science team 2021) is a partnership led by NASA/JPL between US scientists and engineers, with strong collaborations and contributions of the

German, Italian, and French Space Agencies. On 2021 June 2 NASA definitively selected VERITAS as one of the two winners of the Discovery 2019 competition. The launch is expected in the 2028–2030 time frame.

A key scientific objective of VERITAS is understanding the links between the interior, surface, and atmospheric evolution. The determination of the tidal response, tidal phase lag, and MOIF is specifically focused on pushing forward our understanding of the Venus interior. VERITAS will carry two science instruments: VISAR (Venus Interferometric Synthetic Aperture Radar), the X-band interferometric radar (Hensley et al. 2020); and VEM (Venus Emissivity Mapper), an infrared spectroscopic mapper (Helbert et al. 2020). Data from VISAR will be combined with two-way dual X- and Ka-band Doppler tracking data provided by the onboard telecom subsystem collected for the gravity science investigation and used to improve estimates of the Love number k_2 , the tidal phase lag δ_{k_2} , and the MOIF in order to constrain the structure of the Venus interior.

Arriving at Venus after a 6-month cruise, VERITAS will begin an 11-month aerobraking phase, paused after 5 months for 5 months of VEM science observations, before continuing to its final nearly circular polar orbit (180×255 km in altitude, $\sim 85.4^\circ$ inclination, period ~ 1.5 hr). VERITAS plans to operate for 4 Venus sidereal days (or four cycles, 243 Earth days each), providing a nearly global coverage of the planet for all its investigations (gravity science, VISAR, and VEM).

The goal of this work is to simulate the operational scenario of VERITAS's gravity experiment to assess the accuracy in the estimate of k_2 , δ_{k_2} , and MOIF. In our work, alongside a typical orbit determination solution employing Earth Doppler tracking data, we explore a novel approach based on the systematic inclusion of VISAR landmark features observations, or tie points, to tighten the determination of the rotational state of the planet. The simulations presented here, moreover, represent the first assessment of the impact of recent advancements in the



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understanding of Venus’s atmospheric dynamics, namely, the gravitational signature of atmospheric tides (Bills et al. 2020) and short-term sidereal period oscillations of the solid planet due to the transfer of atmospheric angular momentum (Margot et al. 2021).

In Section 2 we describe the concept and the assumptions used in our simulations for both Doppler and radar measurements (Sections 2.1 and 2.2, respectively) and their combination (Section 2.3). In Section 3 we discuss the simulation setup and observational scenario. In Section 4 we present and discuss the results of the simulations. Section 5 follows with concluding remarks. This manuscript is complemented by six appendices.

2. Data and Methods

It is well known that the sole knowledge of the gravitational field is not enough to infer the moments of inertia of a planet, which provide crucial constraints on its interior structure. To constrain the inertia tensor of a body, the gravity field information must be complemented by measurements of the rotational state. Precise Doppler tracking data, the primary observable quantity for gravity field recovery, are quite sensitive to the rotational state of the planet, but the attainable accuracy can be improved by augmenting the analysis with surface feature tracking. The latter provides direct observations of the rotational motion of the planet by measuring the inertial displacement of physical features located on the planet’s surface. In this work, we make use of a novel approach (building on the technique proposed by Davies et al. 1992; Chodas et al. 1992) to combine Earth-spacecraft Doppler tracking data and repeated surface landmark observations (tie points) provided by the onboard interferometric SAR.

2.1. Spacecraft Doppler Tracking

Doppler measurements are the primary observables for reconstructing the orbit of the spacecraft and recovering the gravity field of a planet. These measurements are collected by recording the Doppler shift of a radio signal sent from the ground station to the spacecraft, which then coherently retransmits it back to Earth by means of an onboard transponder (two-way configuration). VERITAS’s Doppler tracking system, with its heritage from ESA’s BepiColombo (Iess et al. 2009, 2021), is able to establish two simultaneous coherent radio links in the X (7.1–8.5 GHz) and Ka (31.8–34.7 GHz) bands and to provide measurements of the range rate of the probe with an average accuracy of 0.01 mm s^{-1} (Ka band, 60 s integration time) under nominal operational conditions (Cappuccio et al. 2020). The dual-link configuration can be used near superior solar conjunctions to suppress about 75% of the noise due to charged particles in the solar corona (Bertotti et al. 1993). In addition, the tracking system is capable of range measurements at the level of 2–3 cm at Ka band (Cappuccio et al. 2020).

The operational scenario of VERITAS consists of five to seven Doppler tracking passes a week, collected by NASA’s Deep Space Network (DSN) ground stations. The VERITAS observation schedule entails approximately a daily contact to ground for 8 hr, as well as 16 hr of VISAR observations. For the gravity experiment we simulate 8 hr passes for 5 days a week collected by DSN station DSS 25 (Goldstone, CA). The integration time of the Doppler observables is set to 10 s, corresponding to a displacement of the spacecraft of $\sim 70 \text{ km}$

on the surface, which is sufficient to resolve gravity field features as small as 190 km after four cycles or $l > 100$ globally (the high-resolution gravity mapping results are beyond the scope of this work and will be published separately; for preliminary results refer to Mazarico et al. 2019).

2.2. Radar Observations and Tie Points

VISAR is an X-band interferometric radar operating at 7.9 GHz (3.8 cm wavelength) and has a 20 MHz bandwidth from which radar imagery with 30 m ground resolution pixels and topographic data with 250 m spatial resolution and 5 m elevation accuracy is produced. The radar acquires data with a look angle of 30° (angle between the antenna boresight and spacecraft nadir) and images a swath width of 14.4 km (for the VISAR flight configuration and observing geometry see Figure 1 in Hensley et al. 2020). After an orbital period, planetary rotation shifts the VERITAS ground tracks by $\sim 10 \text{ km}$; thus, the swath width provides more than 2 km of overlap between swaths acquired on adjacent orbits, enabling coherent mapping of Venus’s surface. Radar data are collected on 11 out of 16 orbits per day and downlinked to Earth on the remaining 5 orbits, when two-way X- and Ka-band tracking data are acquired. Therefore, radar and Doppler data are not collected simultaneously.

VISAR transmits pulses and records the received echoes to generate images of the backscatter signal from the surface. To achieve fine resolution in the radar along-track or azimuth direction, SAR image formation combines echoes from multiple pulses when a point is illuminated by the radar antenna beam. The pixel location in a radar image is determined by the range, i.e., distance from the platform to the pixel, and the Doppler frequency, i.e., projection of spacecraft velocity on the line of sight. For Venus, the range, derived from the delay between pulse transmit time and echo return time, must be corrected for the delay induced by the thick Venus atmosphere. Atmospheric contribution amounts to 200–400 m of additional range, depending on the pixel elevation and imaging geometry. Since VISAR is an interferometer, it solves for the three-dimensional position of each pixel using the range, Doppler, and interferometric phase from two spatially separated antennas.

Surface features (landmarks) imaged on multiple orbits can be identified using automated matching software. The relative range and Doppler measurement errors depend on how accurately imagery acquired from different orbits can be matched. SAR image matching is hindered by speckle that results in a grainy appearance due to the coherent nature of imaging and from differences in imaging geometry, either incidence angle or look direction. Matching accuracy is a function of the image signal-to-noise ratio, the number of looks used to reduce both speckle and thermal noise, imaging geometry differences, and the amount of scene contrast (see Appendix A.1).

Identification of radar tie points will use an automated scene matching algorithm. The automated matching algorithm computes the cross-correlation for a search window that covers the largest expected offset due to ephemeris errors. To account for the spatially variable nature of the matching accuracy and the consequent range and Doppler measurement error, we adopt the match covariance matrix used in the automated matching algorithm to estimate the matching accuracy (Frankot et al. 1994). We tune the matching metric based on match

accuracy statistics from Magellan stereo data that covered approximately 20% of the surface (see Appendix A.1).

The average accuracy of the range and Doppler observations of each radar tie point is 3 m and 10 Hz, respectively, derived from an average 0.2 pixel matching accuracy using a 32×32 matching window of 30 m resolution pixel imagery, where each pixel corresponds to 15 m of range and 40 Hz of Doppler.

To include radar tie points in our simulations, we generated a simulated data set of radar observations. Two types of radar tie points were simulated. The first type (local tie points) is observed in the swath overlap region of adjacent orbits. These measurements permit better orbiter trajectory determination by providing constraints between adjacent orbits when VERITAS is not tracked by the DSN. The second type of radar tie points are the so-called global tie points. A point on the surface can in principle be imaged up to eight times (excluding swath overlaps, i.e., local tie points) during the four-cycle mission: one time each on the descending and ascending passes, for each of the four cycles. Each observation is separated by half a Venus sidereal period, thus enabling us to place tight constraints on the inertial motion of surface features, directly related to the rotational state of the planet. The landmarks are defined in the Venus body fixed frame (see Appendix A.2).

For the simulation, we placed landmarks on a latitude/longitude grid with approximately 150 km spacing separating points in both directions. We exclude orbits that are used for data downlink, in solar conjunction or in power-restricted orbits where data are not collected. A total of 967,605 tie points could be obtained from a set of 387,382 unique landmarks, but we apply downsampling in the simulations (Section 3).

2.3. Joint Inversion

Our approach consists of a systematic joint inversion of both Earth Doppler tracking and VISAR tie point data sets for the simultaneous retrieval of gravity, rotation, tidal response of Venus, and the location of the geodetic control network composed by the tracked landmarks. The simultaneous solution for the spacecraft orbit and the landmark positions allows us to place tight constraints between the planetary body fixed frame and the inertial frame, increasing by an order of magnitude the sensitivity (as shown in Section 4) to the rotational state of the planet. Differently from what was done for Magellan (e.g., Davies et al. 1992), in our work we do not solve separately for the spacecraft ephemerides and the landmark positions, or apply the joint inversion to only a limited subset of orbits. For the first time, we implement the joint inversion of the two data sets for the full gravity and rotation solution in a systematic way.

A two-step solution (i.e., the preliminary inversion of the tracking data and the subsequent inversion of the tie points) suffers the unescapable problem of the propagation of the orbital reconstruction errors in the geodetic control network solution, causing systematic errors that are difficult to evaluate and mitigate. The single-step solution adopted in this work, where Doppler data and tie points are jointly processed, overcomes this problem. An important aspect that needs to be emphasized is that with this approach no a priori information regarding the landmark registration accuracy is required. The accuracy of landmark position recovery is directly estimated in the inversion process.

The substantial increase in sensitivity becomes an efficient way to overcome the limitations for future high-precision

gravity and rotation experiments at Venus due to atmospheric tides (Bills et al. 2020) and high variability of the sidereal period, recently observed by Margot et al. (2021). Both these aspects require a precise tying of the spacecraft orbit to the rotational motion of the planet. The tie point inclusion gives robustness to the solution by providing a direct observation of the rotational motion.

3. Numerical Simulations

To assess the capabilities of VERITAS to retrieve the rotational state, Love numbers, and MOIF, we conducted an extensive set of numerical simulations replicating the nominal operational scenario of VERITAS.

We assessed the capabilities of VERITAS through a covariance analysis. Using the JPL orbit determination software MONTE (Evans et al. 2018), we integrate the trajectory of the probe, generate synthetic Doppler and VISAR data according to the assumptions outlined in Sections 2.1 and 2.2, and superimpose white Gaussian noise. To account for the noise variability observed by Cappuccio et al. (2020), we draw the noise level for each arc from a uniform distribution ranging between 0.015 and 0.038 mm s⁻¹ for Earth-spacecraft Doppler tracking. The noise assigned to the radar tie points is 3 m and 10 Hz in range and Doppler, respectively. We then combine all the data in a least-squares filter (ORACLE) developed at Sapienza University and validated with several space missions (e.g., Iess et al. 2018). The filter implements a multiarc approach that is best suited for data analysis of long-duration gravity experiments (e.g., Durante et al. 2020; Mazarico et al. 2014; Konopliv et al. 2013).

We randomly downsampled the full set of simulated landmarks to $\sim 12,000$ and constructed both global and local tie points. The choice of simulating only a subset of landmarks is supported by two arguments. First, observations of a landmark-dense area might be highly correlated. Selecting only well-spaced points supports the assumption that the observations are statistically independent, therefore simplifying the analysis. Second, the outcome of the simulation can be considered a conservative estimate of what would be possible if the entire data set is processed (for a discussion on the influence of the number of measured landmarks, refer to Appendix A.3).

The dynamical model used to propagate the spacecraft trajectory includes the monopole gravitational acceleration of all main solar system bodies, a degree and order 50 static gravity field of Venus (derived from Konopliv et al. 1999, we limit the spherical expansion to degree 50 since higher degrees have negligible effects on the parameters of interest), the tidal response to the Sun, the nongravitational accelerations due to solar radiation pressure and atmospheric drag, and wheel desaturation maneuvers. To account for possible mismodeling of the nongravitational forces, we employ a large set of empirical accelerations with conservative a priori uncertainties (see Appendix A.4).

Our model also includes atmospheric tides, as the spacecraft tracking system will be sensitive to their effect (Goossens et al. 2018; Bills et al. 2020). The numerical results we report in the next paragraphs are based on the assumption of a knowledge of the atmospheric tidal model with 10% accuracy. Atmospheric tide modeling and the effect of the assumed a priori knowledge on the final results are discussed in Appendix A.5.

Table 1
Results of the Numerical Simulations

Parameter	Current Uncertainty from Observations	Source	Earth Doppler Only	Earth Doppler + Tie Points	Tie Point Improvement Factor
α_0 (arcsec)	3	Margot et al. (2021)	2.7	0.26	
δ_0 (arcsec)	2.5		1.5	0.15	
ϵ (arcsec)	3		1.2	0.12	~ 10
Ω (deg century $^{-1}$)	9.2×10^{-2}		3.8×10^{-2}	3.6×10^{-3}	
MOIF	2.4×10^{-2}		1.0×10^{-2}	9.8×10^{-4}	
k_2	6.6×10^{-2}	Konopliv & Yoder (1996)	1.3×10^{-3}	4.6×10^{-4}	~ 3
δ_{k_2} (deg)	1.20×10^{-1}	4.5×10^{-2}	

Note. We report the uncertainties in the two cases of Doppler only and Doppler + tie point analyses (three times the formal uncertainty). The tie point improvement factor is the ratio between the uncertainties obtained without and with the inclusion of VISAR data.

According to recent observation campaigns of the Venus rotation rate (Margot et al. 2021), the complex coupling between the atmosphere and the planet results in sidereal period variations significantly larger than what was predicted by general circulation models (GCMs; Lebonnois et al. 2010; Cottareau et al. 2011), leading to variations of the sidereal period up to ~ 3 and ~ 20 minutes over timescales of 1 Earth day and 117 Earth days, respectively. If not correctly modeled, these variations induce an error in the longitude positioning of surface features that grows in time. We accounted for this perturbation in our simulations by estimating a sidereal period every 2.5 days, setting the conservative a priori uncertainty value of 20 minutes over one arc. For the full set of estimated parameters and the detailed filter setup and assumptions, refer to Appendix A.4.

4. Results and Discussion

Table 1 reports the uncertainties (all results in tables and text are given as three times the formal uncertainty, or 3σ) attainable for the Venus rotational parameters, the Love number, and the MOIF in the nominal VERITAS mission configuration for two cases: Doppler tracking data only, and Doppler tracking data combined with VISAR observations. The inclusion of VISAR tie point measurements in the orbit determination enables a large improvement in the determination of the rotational state of Venus, not attainable with Doppler data alone. The tie points increase the accuracy in the pole location and MOIF by about a factor of 10, while a smaller improvement (~ 3) is found on k_2 and its tidal phase lag.

The current estimate of Venus Love number (0.295 ± 0.066 , 2σ ; Konopliv & Yoder 1996), coupled with the lack of a magnetic field, does not resolve a liquid or solid core (Dumoulin et al. 2017; see Figure 1). The analysis by Dumoulin et al. (2017) indicates that the state of the core and its size, the mantle composition, and the viscous response of the interior, or rather different classes of interior conditions, can be well constrained with a knowledge of k_2 to an accuracy smaller than 3% (0.01, the 1σ VERITAS requirement) and a precise measurement of the phase lag (VERITAS has a 1σ requirement of $\sigma_{\delta_{k_2}} = 0.25^\circ$). In this work we assume the classical definition of the tidal potential phase lag (Murray & Dermott 2000), adopted also in the models by Dumoulin et al. (2017).

Thanks to the augmentation provided by VISAR tie points, our simulations show that VERITAS will be able to determine these tidal quantities with an accuracy substantially better than these threshold values (see Table 1). With the joint processing of radio tracking data and radar tie points, the right ascension and declination of the pole (α_0 and δ_0) can be determined with an accuracy increased by an average factor of 10, improving the results obtained by Magellan by more than 100 times and the ground-based observations (Margot et al. 2021) by more than an order of magnitude. A comparable improvement is found for the obliquity ϵ ($\sigma_\epsilon = 0.12$ arcsec). The considerable improvement with respect to Magellan is mainly due to the more favorable VERITAS orbital geometry, the longer time span of the gravity observations, and the substantial improvements in the end-to-end radio tracking performance (the use of Ka-band, dedicated instrumentation for media calibration, for both charged particles and tropospheric water vapor, and open loop ground receivers). The use of radar tie points also leads to substantial improvements in orbit determination, and hence in the gravity and rotational state recovery (see Table 1). VERITAS will also measure the pole precession rate Ω and derive the MOIF. Assuming that the spin axis of Venus precedes in a conical motion about the orbit normal (as detailed in Appendix A.6), the precession rate can be determined to a level of 3.6×10^{-3} deg century $^{-1}$. The corresponding relative uncertainty in the MOIF is 0.3% of its predicted central value (0.336; Cottareau & Souchay 2009). The accurate measurement of the MOIF provides an additional, strong constraint to models of the Venus interior, by reducing the uncertainty in the density of the core and the mantle, considering the core size as constrained by k_2 (see discussion below). As is well known (see, e.g., Bills & Rubincam 1995), MOIF alone cannot uniquely determine the interior structure, even for a two-layer model of the interior. Nonetheless, it is a crucial constraint to every geophysical model of Venus interior.

Based on current models (Dumoulin et al. 2017), the constraints on k_2 and δ_{k_2} can determine the core state and distinguish different classes of interior conditions. For example, if the core is fluid with an Earth-like composition, the core size can be obtained to within 100 km and average mantle viscosity to within an order of magnitude (see Figure 1). The latter value strongly depends on the temperature distribution and the volatile content in the mantle and therefore provides information about the heat and volatile loss of

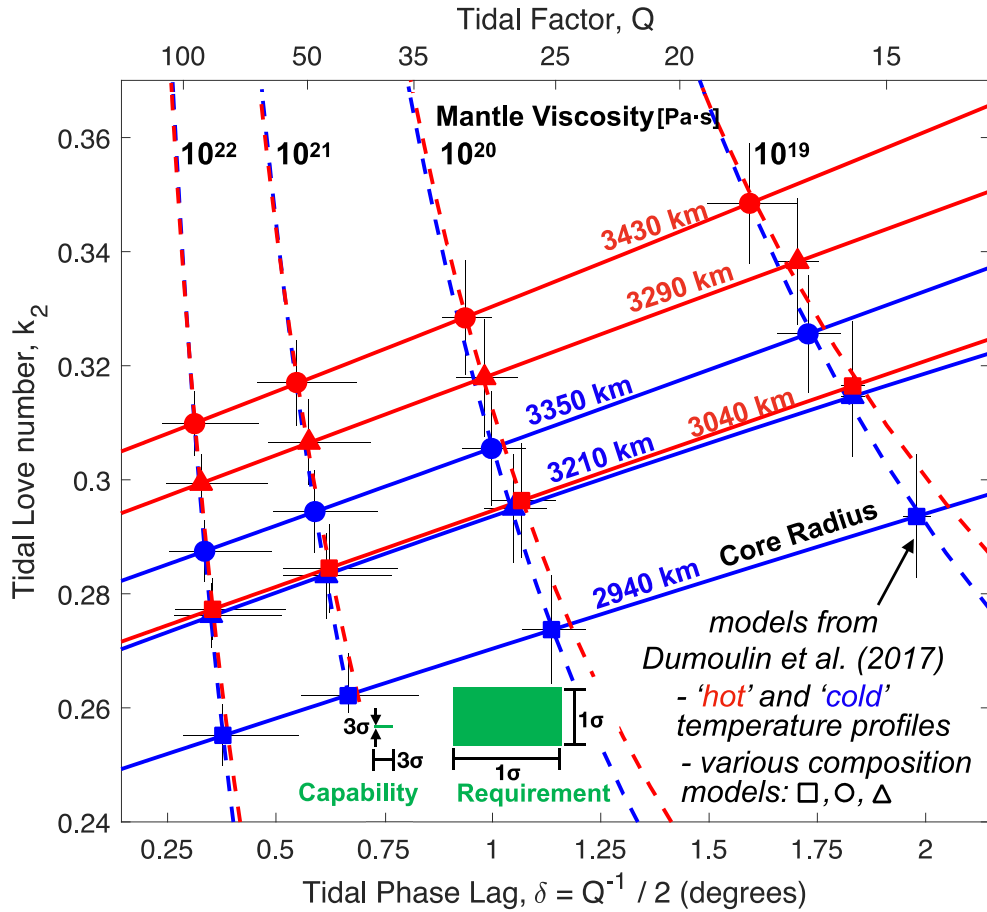


Figure 1. The required uncertainty on Love number k_2 and the tidal phase lag to constrain core radius and mantle viscosity as defined by the VERITAS mission requirements (right green box) is compared with the experiment capability (left green box, here shown at 3 times the formal uncertainty), showing the possibility to constrain the interior structure of Venus in the particular case of a fluid core with Earth-like composition. The black lines represent the error bars of the models by Dumoulin et al. (2017).

the planet. For example, a warm and wet mantle, representative of a planetary interior that has not cooled much and has lost little of its original water, has a low viscosity, while a cold and dry mantle, representative of an efficiently cooled and outgassed interior, has a high viscosity. These two extreme models would differ in viscosity by several orders of magnitude and could be distinguished by the measurement of the phase lag.

Different formation scenarios lead to different compositional models based on cosmochemical assumptions and trends among Earth-like planets to model the interior of Venus. A major difference in the models is the FeO content of the mantle, which can vary between 0.42 and 18.7 wt.%. This results in different values of MOIF ranging between 0.33 and 0.342 (~3% variation), with otherwise the same assumption about the thermal state and the core composition (Dumoulin et al. 2017). In addition to k_2 , knowledge of the MOIF with an accuracy of 0.3% will therefore further help to distinguish the mantle composition models.

The amount of light elements in the core, particularly important for a better understanding of Venus's magnetic field evolution and also informative about Venus's conditions during core formation, is not known. The two parameters together, k_2 and MOIF, help to better distinguish the models as has already been shown, for example, for Mars (Rivoldini et al. 2011; and recently confirmed with InSight seismic data by Stähler et al. 2021), and thus better than in the models of

Dumoulin et al. (2017), for which MOIFs were assumed to be unknown. The information about the density distribution from the MOIF is not unique, i.e., for the same MOIF the core can be small and dense or relatively larger and lighter. If the core of Venus is liquid, the core size can be constrained independently with k_2 and core density can be constrained in combination with the MOIF. The inverse problem of interior structure determination is degenerate, and thus different models can lead to the same values of MOIF, k_2 , and δ_{k_2} . Different models, however, are not equally physically likely, and their selection will rely on additional constraints arising from planetary geology and geophysics as has been done, e.g., on Mars (Rivoldini et al. 2011), Mercury (Genova et al. 2019), or the Saturn satellites (Durante et al. 2019).

All these fundamental quantities, such as core state, size and composition, and mantle composition and viscosity, are necessary to understand the formation of Venus and its thermal and magnetic evolution. They serve, for example, as inputs (core radius and core and mantle composition) or constraints (core state and present effective mantle viscosity) for modeling core and mantle processes and the thermal and magnetic evolution (e.g., O'Rourke et al. 2018).

Additional simulations (Appendix A.5) show that even under the very conservative assumption of an atmospheric tidal model with an uncertainty of 100%, the joint solution approach guarantees the insensitivity of the rotational state solution (and

thus the MOIF) to the atmospheric tide. The tidal quantities (k_2 , δk_2) are still determined within the mission requirements with significant margin (see Table A2).

Based on the recent findings of Margot et al. (2021), and considering the high accuracy of the VERITAS tracking system, a constant sidereal period over the mission duration can no longer be assumed, as was done in the Magellan data analysis (Davies et al. 1992). We tackle the problem of the irregular rotation rate by estimating an average sidereal period every 2.5 days. The uncertainty in the estimated rotation period by processing data acquired in 2.5 days is approximately 9 minutes. Although this uncertainty is larger than the maximum observed variation over 1 Earth day by Margot et al. (2021; about 3 minutes), VERITAS will be able to measure variations of the sidereal period over longer timescales. Indeed, the uncertainty in the sidereal period is with good approximation inversely proportional to the time span of the observations. With the aforementioned retrieved uncertainty, VERITAS would be able to distinguish 3-minute variations over 8 days. Variations of 20 minutes over 117 days (Margot et al. 2021) will be measured to good accuracy by VERITAS. The joint inversion approach proves to be extremely valuable here as well: with Doppler data only, the achievable accuracy of the 2.5-day-averaged sidereal period increases to 43 minutes. We point out that although short-term variations of the sidereal period (length of day) result in variations of the longitudinal position of the landmarks, the spin axis solution is robust. If $r_i(t)$ is the position of a landmark in the Venus-fixed, Venus-centered frame at time t , then the spin axis $\hat{\omega}$ is determined by the orthogonality condition $\hat{\omega} \cdot (r_i(t_n) - r_i(t_m)) = 0$. Any small error in the longitude of the landmarks does not affect the orthogonality condition, in the presence of a slow precession. Thus, it is only the latitudinal position of the landmarks that affects the determination of the pole position.

As a by-product of the estimation process, we can refine the location of all observed landmarks and in particular the ones contributing to the *global* tie points, thus providing the backbone of an accurate geodetic control network. The median values of the recovered global landmark position accuracy in altitude, latitude, and longitude (mapped on the reference surface of Venus) are $M_{\text{alt}} = 3 \text{ m}$, $M_{\text{lat}} = 7 \text{ m}$, and $M_{\text{lon}} = 6 \text{ m}$, respectively. The determination of the landmark positions also enables the retrieval of the radial displacement associated with the tidal forcing, parameterized by the Love number h_2 . This tidal Love number has so far only been measured from orbit only for the Moon (Mazarico et al. 2014) and Mercury (Bertone et al. 2021). The retrieved uncertainty for VERITAS ($\sigma_{h_2} = 0.15$, equivalent to $\sim 10 \text{ cm}$ of maximum radial displacement) corresponds to a relative uncertainty on h_2 (predicted value of 0.45–0.75; Dumoulin et al. 2017) of 20%–33%. While the relative uncertainty is not sufficient to constrain interior models, it will serve as an additional check.

The inclusion of tie points improves the orbital solution by providing observability during periods in which the spacecraft is not tracked from Earth. This aspect is particularly important for providing a uniform positional coverage along the orbit and giving robustness to the determination of physical effects such as atmospheric drag, whose variability cannot be predicted with enough accuracy by a deterministic a priori model. The denser availability of orbit-related data is also the reason for a slight improvement in the retrieval of the low-degree gravity field, since the nearly continuous orbital coverage allows a better resolution of the large spatial scales (i.e., low degree) of the planetary gravitational field.

5. Conclusions

By simulating the nominal mission scenario, we show that the VERITAS mission to Venus, planned for launch in 2027–2028, has the capability to determine crucial parameters (tidal Love number k_2 and MOIF) needed to substantially improve models of the planet’s interior structure. The precise characterization of the tidal response of Venus via the measurement of its complex Love number will allow us to place improved constraints on the state and size of the core and on the viscous response of the planet to tidal stresses, such that different classes of interior conditions can be obtained. We show that our data analysis approach, where Doppler tracking data and radar tie points are jointly processed, is extremely effective in the determination of the rotational state of the planet and the moment of inertia factor. It pushes further the possibility of understanding of the dynamical evolution of Earth’s neighboring planet. In particular, the determination of k_2 and MOIF together will help better constrain the core size, as well as the core and mantle composition. These constraints are the first-order information necessary to address the question of why Venus lacks a dynamo. Models of interior structure, temperature, and composition compatible with the measurements (k_2 , δk_2 , and MOIF) would provide the present-day boundary condition for thermal evolution models of Venus. The current dearth of information on these fundamental characteristics of Venus’s interior precludes meaningful comparisons between the different evolutionary paths of Venus and the other bodies of the inner solar system. Understanding the present-day state of Venus’s interior and its past evolution will offer valuable clues as to how and why Venus evolved into an uninhabitable planet.

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Appendix

A.1. Tie Points Simulation

The match covariance matrix is given by

$$M_{\text{cov}} = k_c H^{-1} \left(2\sigma_n^2 H + \frac{1}{2} A_w \sigma_n^4 I \right) H^{-1}, \quad (\text{A1.1})$$

where k_c is an empirical constant inferred from Magellan match statistics, I is the identity matrix, and H is the Hessian of the match correlation function $c(x, y)$. For a given image offset (x, y) the Hessian is given by

$$H = \begin{bmatrix} \frac{\partial^2 c}{\partial x^2} & \frac{\partial^2 c}{\partial x \partial y} \\ \frac{\partial^2 c}{\partial x \partial y} & \frac{\partial^2 c}{\partial y^2} \end{bmatrix}, \quad (\text{A1.2})$$

A_w is the area in pixels of the matching window, and σ_n is a measure of the backscatterer difference between the two images in the matching window

$$\sigma_n = \frac{1}{2A_w} \sum_{x \in w} [I_1(\vec{x}) - \bar{I}_1 - I_2(\vec{x} - \vec{o}) + \bar{I}_2(\vec{o})]^2, \quad (\text{A1.3})$$

where $I_1(\vec{x})$ and $I_2(\vec{x})$ are the pixel intensities for the two images at position \vec{x} , \vec{o} is the offset vector between the images, and \bar{I}_1 and $\bar{I}_2(\vec{o})$ are mean intensities in the match window. We approximate the correlation function for a good match by the product of sinc functions given by

$$c(x, y) = \text{sinc}\left[\frac{\pi x}{2\sigma_{m_x}}\right] \text{sinc}\left[\frac{\pi y}{2\sigma_{m_y}}\right], \quad (\text{A1.4})$$

where σ_{m_x} and σ_{m_y} are the matching accuracy in pixel in the x - and y -directions. Differentiating Equation (A1.4) twice and evaluating at the peak yields

$$H = \begin{bmatrix} -\frac{1}{12} \frac{\pi^2}{\sigma_{m_x}^2} & 0 \\ 0 & -\frac{1}{12} \frac{\pi^2}{\sigma_{m_y}^2} \end{bmatrix}, \quad (\text{A1.5})$$

where σ_{m_x} and σ_{m_y} are given by

$$\sigma_{m_q} = \frac{1}{2} - k_{m_q} \frac{\sigma_{10 \times 10}}{\bar{\sigma}_{10 \times 10} \frac{1}{\sqrt{N_L}} \left(1 + \frac{1}{S/N}\right)}, \quad (\text{A1.6})$$

where $q = x, y$, k_{m_q} are empirical parameters based on Magellan match statistics, S/N is the signal-to-noise ratio, N_L are the number of looks (number of single look pixel intensities averaged together in a 30 m multi-looked pixel to reduce thermal and speckle noise), $\bar{\sigma}_{10 \times 10}$ and $\sigma_{\sigma_{10 \times 10}}$ are the mean and standard deviation of the backscatter in a 10×10 pixel window centered at the match point obtained from Magellan imagery, and

$$S/N = \frac{\bar{\sigma}_{10 \times 10}}{\text{NES0}}, \quad (\text{A1.7})$$

where NES0 is the radar noise equivalent sigma naught (backscatter value where the S/N equals 1). To obtain an approximate value for σ_n^2 , we use

$$\sigma_n^2 = \left[\sqrt{2} \bar{\sigma}_{10 \times 10} \left(\frac{1}{\sqrt{N_L}} \frac{1}{S/N} + k_s \frac{\bar{\sigma}_v}{\sigma_{10 \times 10}} \right) \right], \quad (\text{A1.8})$$

where k_s is an empirical value derived from Magellan match statistics and $\bar{\sigma}_v$ is the mean X-band backscatter value for Venus, roughly -10.5 dB.

A.2. Uncertainty on Venus's MOIF from Pole Precession Measurements

The torque of the Sun on Venus determines the precession of its spin axis in a conical motion about the orbit normal. The precession rate Ω is

$$\Omega = \frac{3}{2} \frac{n^2}{\omega} \cos \epsilon \frac{J_2}{k}, \quad (\text{A2.1})$$

where J_2 is the unnormalized degree 2 zonal coefficient of the gravity field of Venus, k is the MOIF, ω is the sidereal spin rate, n is the mean motion, and ϵ is the obliquity (angle between spin axis and orbit normal). If a measurement of the precession rate of the spin axis is available, Equation (A2.1) can be used to estimate the MOIF of the planet. The precession of Venus, deduced from Equation (A2.1), has a period of $\sim 29,000$ yr (Cottareau & Souhay 2009) for reasonable ranges of MOIF

values. Although Ω is relatively large (during the four Venus cycles spanned by VERITAS the angular displacement of the pole is about 0.03°) compared to the small axial tilt of the planet (2.64°), the angular displacement of the pole mapped on Venus's surface corresponds to only ~ 150 m. Note that the attainable accuracy on the MOIF primarily results from the accuracy of the precession rate, since J_2 and the other quantities in Equation (A2.1) are much better known.

As the precession period is much longer than the VERITAS observations, the precessional motion of Venus can be described by three first-degree polynomials for the spin vector right ascension, $\alpha(t) = \alpha_0 + \dot{\alpha}(t - t_0)$, declination, $\delta(t) = \delta_0 + \dot{\delta}(t - t_0)$, and prime meridian, $w(t) = w_0 + \frac{2\pi(t - t_0)}{T_{\text{sid}}}$, with t_0 corresponding to the J2000 epoch and T_{sid} being the sidereal period (see, e.g., Archinal et al. 2011). Due to the recent findings of Margot et al. (2021), a constant T_{sid} cannot be employed, as its short-term variations would induce significant latitudinal variations of the position of the observed landmarks. Indeed, we estimate one T_{sid} for each arc and compute $w(t)$ as follows:

$$w_i(t) = w_{i-1}(t) + \frac{2\pi(t - t_{0i})}{T_{\text{sid}i}},$$

where $w_i(t)$ is the prime meridian expression valid for the i th arc, t_{0i} is the starting epoch of the i th arc, and $T_{\text{sid}i}$ is the sidereal period estimated at the i th arc.

The precession constant Ω can be directly associated (under the assumption that nutations are negligible and small deviation from the reference position, as shown and justified in Appendix A.6) with $\dot{\alpha}$ and $\dot{\delta}$ as

$$\Omega = c_1 \dot{\alpha} = c_2 \dot{\delta}, \quad (\text{A2.2})$$

with the coefficients c_1 and c_2 determined by the orbital inclination of Venus, the reference position of the pole (α_0, δ_0), and the longitude of the orbital node at the reference epoch. In our simulations we therefore estimate the pole polynomial coefficients and exploit the aforementioned procedure to assess the uncertainty on Ω and thus on the MOIF.

A.3. Effect of the Number of Landmarks

To explore the effectiveness of the inclusion of radar tie points, we performed a sensitivity analysis of the results to the number of included landmarks.

We analyzed the formal uncertainty improvement factor as a function of the number of observed landmarks. We ran simulations covering the range of 1000–8000 landmarks. Not surprisingly, the improvement factor P depends on the number of landmarks n as

$$P(n) \sim n^{\frac{1}{2}},$$

a consequence of the assumption that the measurements are statistically independent. The results that we report can be easily scaled to an arbitrarily higher number of landmarks.

The increase in the accuracy of the rotational parameters, MOIF, and k_2 shows that, while the bulk of the information matrix comes from radio tracking data, tie points, being a largely independent data set, increase the overall information content by a quite significant amount.

Table A1
Filter Assumptions

Parameter(s)	A Priori Uncertainty	Comment
Local Parameters		
Orbiter position	0.1 km	
Orbiter velocity	$10^{-5} \text{ km s}^{-1}$	
Solar pressure scale factor	25%	One per arc
Periodic accelerations	$6 \times 10^{-11} \text{ km s}^{-2}$	Corresponding to 25% Cd (drag coefficient) uncertainty; time update: 2 hr
Sidereal period	20 minutes	Corresponding to 15 yr maximum variability observed by Margot et al. (2021)
Wheel desaturation maneuvers	6 mm s^{-1}	
Local landmark positions	Unconstrained	
Global Parameters		
Gravity field spherical harmonics coefficients	Unconstrained	
Gravity field correction coefficients for atmospheric tides	10/50/100%	As detailed in Appendix A.5
Pole right ascension/declination	Unconstrained	
Pole right ascension/declination rate	Unconstrained	
Complex love number	Unconstrained	
Love number h_2	Unconstrained	
Global landmark positions	Unconstrained	

A.4. Filter Setup

We report here the detailed setup of the orbit determination filter used in this work.

The complete list of parameters estimated in the filter includes the following:

State of the orbiter (position and velocity), degree and order 50 gravity field spherical harmonics coefficients, complex Love number k_2 , atmospheric tide parameters as described in Appendix A.5, position (latitude, longitude, radius) of all observed landmarks, Venus pole location (right ascension, declination) and its precession rate, and Venus sidereal period.

Atmospheric density variations, depending on local time and solar activity (Müller-Wodarg et al. 2016; Kliore et al. 1992), induce accelerations on the spacecraft over typical timescales ranging from half to a quarter of the orbital period T . We account for this possible mismodeling by estimating cosinusoidal along-track accelerations with period T and $T/2$ and a time update of $2h$ with an a priori uncertainty corresponding to a 25% error on Cd. To account for possible misrepresentation of the solar radiation pressure, we estimate one scale factor per arc with an a priori uncertainty set to 25%. Moreover, we include the estimation of daily momentum desaturation maneuvers with an a priori uncertainty set to 6 mm s^{-1} in compliance with navigation assumptions (Wallace et al. 2019).

We run the described filter solving for the set of parameters of interest. In our multiarc approach, the tracking data are subdivided into 2.5-day arcs and the parameters are divided into two sets: local parameters (those affecting a single arc, e.g., position and velocity of the orbiter) and global parameters (parameters affecting all the arcs, e.g., the gravity field of the planet). The total number of global parameters amounts to 27,320.

In Table A1 we report the a priori uncertainty assumptions of the filter.

A.5. Atmospheric Tides Modeling

In the dynamical model used in our simulations we included the effect of atmospheric tides. As shown by Bills et al. (2020), the mass transport induced by solar heating of the atmosphere is not a negligible factor for high-precision radio science experiments at Venus. For a realistic assessment of the attainable accuracies of VERITAS, we modeled the time-variable gravity field induced by solar -heating-driven pressure variations of the atmosphere.

The spherical harmonics expansion of the total (static plus atmosphere) gravity field can be written as a function of time t as

$$C_{lm}(t) = C_{lm}^S + \Delta C_{lm}(t). \quad (\text{A5.1})$$

C_{lm} is the total C coefficient of degree l and order m of the gravity field, C_{lm}^S is the static coefficient, and $\Delta C_{lm}(t)$ is the correction due to the time-variable mass transport (the same formulation applies for S_{lm} coefficients, here omitted for brevity).

To determine the time-variable atmospheric contribution, we employed the model developed by Garate-Lopez & Lebonnois (2018) for retrieving surface pressure variations and then converted these perturbations in the associated gravity field coefficient with a technique that includes the atmospheric loading contribution on the solid planet, originally developed for Earth (Petrov 2004) and applied also on Mars (Genova et al. 2016). This procedure produces the time series of spherical harmonics expansions of the atmospheric gravity field. The gravity field perturbation induced by solar heating is a periodic signal of fundamental frequency f_1 , equal to the main forcing

Table A2Results Comparison (in Terms of Formal Uncertainties, 3σ) for Different Levels of A Priori Knowledge of the Atmospheric Thermal Tides Model Parameters

Parameter	10%		50%		100%	
	Earth Doppler Only	Tie Points	Earth Doppler Only	Tie Points	Earth Doppler Only	Tie Points
α_0 (arcsec)	2.7	0.26	3.0	0.27	3.2	0.27
δ_0 (arcsec)	1.5	0.15	1.7	0.16	1.8	0.16
ϵ (arcsec)	1.2	0.12	1.3	0.12	1.4	0.12
Ω (deg century $^{-1}$)	3.8×10^{-2}	3.6×10^{-3}	4.1×10^{-2}	3.7×10^{-3}	4.4×10^{-2}	3.7×10^{-3}
MOIF	1.0×10^{-2}	9.8×10^{-4}	1.1×10^{-2}	1.0×10^{-3}	1.2×10^{-2}	1.0×10^{-3}
k_2	1.3×10^{-3}	4.6×10^{-4}	1.8×10^{-3}	9.9×10^{-4}	2.3×10^{-3}	1.5×10^{-3}
δ_{k_2} (deg)	1.20×10^{-1}	4.5×10^{-2}	1.9×10^{-1}	1.4×10^{-1}	2.9×10^{-1}	2.5×10^{-1}

effect (i.e., Venus solar day ~ 117 days). We isolated the contribution of the fundamental frequency and its three first harmonics via a Fourier transform. Thus, we can expand Equation (A5.1) as

$$C_{lm}(t) = C_{lm}^S + \sum_n \Delta C_{lm}^n(t) \quad (\text{A5.2})$$

$$\Delta C_{lm}^n = A_{C_{lm}}^n \cos(2\pi f_n t) + B_{C_{lm}}^n \sin(2\pi f_n t), \quad (\text{A5.3})$$

where $f_n = nf_0$ with $n = 1, 2, 3, 4$ and A, B are coefficients derived from the Fourier analysis specific for each coefficient, degree, and order.

In our simulations we assessed the necessity of including these effects in the dynamical model of VERITAS, as its extremely precise tracking system is sensitive to the main components of the thermal tide perturbation. In particular, we assessed that if thermal tides are not accounted for, significant biases might arise in the gravity field and rotational state solution, in particular affecting the Love number k_2 . The most recent analysis of the Venus gravity field accounted for the atmospheric contribution by forward-modeling its effect (Goossens et al. 2017, 2018). We have chosen to adopt a conservative approach and account for the intrinsic uncertainty of the atmospheric tide modeling. We model the thermal tide field up to the degree and order that guarantees that the higher degrees produce no residual signal in the Doppler residuals (i.e., degree and order 18 for f_1 , 13 for f_2 , 7 for f_3 , and 10 for f_4) and considered the uncertainty associated with the correction coefficients $A_{C_{lm}}^f, A_{S_{lm}}^f, B_{C_{lm}}^f, B_{S_{lm}}^f$ for the frequencies f_1 through f_4 .

We evaluated the effect of the assumed a priori knowledge of the atmospheric tide model, without delving into a detailed analysis of atmospheric-dynamics-related sources of uncertainty. We have chosen, then, to assume a certain level of uncertainty on the output of the model, i.e., the correction coefficients. In particular, we explored three cases by setting different a priori uncertainties. We considered an accurate model (model uncertainty equal to 10%), a medium-accuracy model (50% uncertainty), and a coarse-accuracy model (100% uncertainty). In Table A2 we report the results relative to each of the three assumptions. It is important to note how the results, when combining Doppler data and tie points, become significantly less sensitive to the accuracy of the model, for all the parameters except k_2 and δ_{k_2} , which, not surprisingly, have significant sensitivity to the atmospheric tides. This indicates that even a coarse a priori knowledge of the model is sufficient to meet the scientific objectives of VERITAS.

A.6. Relating Precession and Pole Coordinates

In this section we will obtain the equations to express the motion of the pole as a function of the equatorial coordinates

and their time derivatives. This relation has been used in the simulations as a constraint in the determination of the precession rate and the MOIF. Finally, we will show that the errors committed by neglecting the nutations of the pole have negligible consequences in the determination of the precession rate.

The Venus ecliptic (V_E) and the (usual) Earth ecliptic (E_E) reference frames are represented by the unit vectors $\{u_{V,x}, u_{V,y}, u_{V,z}\}$ and $\{u_{E,x}, u_{E,y}, u_{E,z}\}$, respectively. The equatorial frame is represented by $\{u_{eq,x}, u_{eq,y}, u_{eq,z}\}$.

We will use the following coordinates:

1. $\alpha(t), \delta(t)$ are the right ascension and declination (equatorial J2000 coordinates);
2. $\lambda(t), \beta(t)$ are ecliptic coordinates referred to the E_E reference frame at J2000.0;
3. $\lambda_V(t), \beta_V(t)$ are ecliptic coordinates referred to the V_E reference frame at J2000.0

We define

1. the direction (as a unit vector) P_V of Venus's pole;
2. the direction (as a unit vector) P_{0V} of the normal to the Venus orbital plane (hereafter the "orbital pole").

All coordinates above will be referred to the pole position. $\beta_V(t)$ is the nutation in obliquity, and $\lambda_V(t)$ is the sum of the precession and the nutation in longitude.

The direction P_V in the three reference frames is

$$P_{V,eq} = \cos[\delta(t)] \cos[\alpha(t)] u_{eq,x} + \cos[\delta(t)] \sin[\alpha(t)] u_{eq,y} + \sin[\delta(t)] u_{eq,z} \quad (\text{A6.1})$$

$$P_{V,E_E} = \cos[\beta(t)] \cos[\lambda(t)] u_{E,x} + \cos[\beta(t)] \sin[\lambda(t)] u_{E,y} + \sin[\beta(t)] u_{E,z} \quad (\text{A6.2})$$

$$P_{V,V_E} = \cos[\beta_V(t)] \cos[\lambda_V(t)] u_{V,x} + \cos[\beta_V(t)] \sin[\lambda_V(t)] u_{V,y} + \sin[\beta_V(t)] u_{V,z}, \quad (\text{A6.3})$$

while the orbital pole direction in the E_E frame is

$$P_{0V,E_E} = \sin i_0 \sin \Omega_0 u_{E,x} - \sin i_0 \cos \Omega_0 u_{E,y} + \cos i_0 u_{E,z}, \quad (\text{A6.4})$$

where $i_0 = 3.39466189^\circ$ (inclination) and $\Omega_0 = 76.67992019^\circ$ (longitude of the ascending node), at J2000.0 (Simon et al. 1994).

The transformations of P_V from equatorial to E_E coordinates (and vice versa) are

$$P_{V,E_E} = R^{-1} P_{V,eq} \text{ and } P_{V,eq} = R P_{V,E_E}, \quad (\text{A6.5})$$

where

$$R = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \epsilon & -\sin \epsilon \\ 0 & \sin \epsilon & \cos \epsilon \end{pmatrix} \quad (\text{A6.6})$$

and $\epsilon = 23.43662$ is Earth's obliquity.

Equation (A6.5) corresponds to

$$\cos \beta \cos \lambda = \cos \alpha \cos \delta \quad (\text{A6.7})$$

$$\cos \beta \sin \lambda = \cos \epsilon \cos \delta \sin \alpha + \sin \epsilon \sin \delta \quad (\text{A6.8})$$

$$\sin \beta = \cos \epsilon \sin \delta - \sin \epsilon \cos \delta \sin \alpha. \quad (\text{A6.9})$$

At J2000.0 ($t = 0$) the equatorial coordinates of the pole of Venus are $\alpha_0 = \alpha(0) = 272.76^\circ$ and $\delta_0 = \delta(0) = 67.16^\circ$ (Archinal et al. 2011).

By solving Equations (A6.7)–(A6.9), we obtain the pole position at the same epoch in E_E coordinates:

$$\lambda_0 = 30.079869^\circ, \beta_0 = 88.762332^\circ. \quad (\text{A6.10})$$

The $\mathbf{u}_{V,x}$, $\mathbf{u}_{V,y}$, $\mathbf{u}_{V,z}$ directions of the V_E reference frame are as follows:

1. The z -axis points toward the orbital pole $\mathbf{u}_{V,z} = P_{0V}$.
2. The x -axis is the direction of the vernal equinox of Venus: the vernal equinox of Venus coincides with the coordinates of the ascending node of the orbit of Venus at J2000.0 with respect to the equator of Venus at the same date, so $\mathbf{u}_{V,x} = P_{0V} \times P_V$.

From Equations (A6.2) and (A6.4) we obtain

$$\mathbf{u}_{V,x} = u_{V,x1}\mathbf{u}_{E,x} + u_{V,x2}\mathbf{u}_{E,y} + u_{V,x3}\mathbf{u}_{E,z}, \quad (\text{A6.11})$$

where

$$u_{V,x1} = \frac{\cos \beta_0 \cos i_0 \sin \lambda_0 + \sin \beta_0 \sin i_0 \cos \Omega_0}{\sqrt{1 - [\sin \beta_0 \cos i_0 - \cos \beta_0 \sin i_0 \cos (\lambda_0 - \Omega_0)]^2}} \quad (\text{A6.12})$$

$$u_{V,x2} = \frac{\sin \beta_0 \sin i_0 \sin \Omega_0 - \cos \beta_0 \cos i_0 \cos \lambda_0}{\sqrt{1 - [\sin \beta_0 \cos i_0 - \cos \beta_0 \sin i_0 \cos (\lambda_0 - \Omega_0)]^2}} \quad (\text{A6.13})$$

$$u_{V,x3} = -\frac{\cos \beta_0 \sin i_0 \cos (\lambda_0 - \Omega_0)}{\sqrt{1 - [\sin \beta_0 \cos i_0 - \cos \beta_0 \sin i_0 \cos (\lambda_0 - \Omega_0)]^2}} \quad (\text{A6.14})$$

and

$$\mathbf{u}_{V,z} = u_{V,z1}\mathbf{u}_{E,x} + u_{V,z2}\mathbf{u}_{E,y} + u_{V,z3}\mathbf{u}_{E,z}, \quad (\text{A6.15})$$

with

$$\begin{aligned} u_{V,z1} &= \sin i_0 \sin \Omega_0; u_{V,z2} \\ &= -\sin i_0 \cos \Omega_0; u_{V,z3} = \cos i_0, \end{aligned} \quad (\text{A6.16})$$

and finally, $\mathbf{u}_{V,y} = \mathbf{u}_{V,z} \times \mathbf{u}_{V,x}$.

We define M (at J2000.0) as

$$\begin{aligned} M &= \begin{pmatrix} u_{V,x1} & u_{V,y1} & u_{V,z1} \\ u_{V,x2} & u_{V,y2} & u_{V,z2} \\ u_{V,x3} & u_{V,y3} & u_{V,z3} \end{pmatrix} \\ &\approx \begin{pmatrix} 0.531509 & -0.84509 & 0.0576204 \\ 0.846837 & 0.531678 & -0.0136422 \\ -0.0191066 & 0.0560461 & 0.998245 \end{pmatrix}. \end{aligned} \quad (\text{A6.17})$$

The $\alpha(t)$ and $\delta(t)$ coordinates as functions of $\beta_V(t)$, λ_V are given by the following relations:

$$P_{V,\text{eq}}(t) = RMP_{V,V_E}(t)P_{V,V_E}(t) = M^{-1}R^{-1}P_{V,\text{eq}}(t). \quad (\text{A6.18})$$

The pole motion around the orbit pole is ($t = 0$ corresponds to J2000.0)

$$\lambda_V(t) = \lambda_V(0) + \Omega t + \delta\lambda_V(t) \quad (\text{A6.19})$$

$$\beta_V(t) = \beta_V(0) + \delta\beta_V(t), \quad (\text{A6.20})$$

where $\delta\beta_V$, $\delta\lambda_V$ are the nutations in obliquity and in longitude, respectively, and Ω is the precession rate of the Venus pole. The precession rate is the sum of the solar precession ($\sim 44.74 \text{ yr}^{-1}$) and the planetary precession ($-10'' \text{ yr}^{-1}$; Simon et al. 1994). By solving for β_V in Equation (A6.18), we get

$$\beta_V(0) = 87.3638\lambda_V(0) = 90^\circ. \quad (\text{A6.21})$$

By differentiating Equation (A6.18) with respect to time and neglecting the nutations, we get

$$\begin{aligned} \frac{\dot{\alpha}(t)}{\delta(t)} &= \frac{\sin \alpha(t) \tan \delta(t) (\cos \epsilon \sin i_0 \cos \Omega_0 + \sin \epsilon \cos i_0)}{\cos \alpha(t) (\cos \epsilon \sin i_0 \cos \Omega_0 + \sin \epsilon \cos i_0) + \sin i_0 \sin \Omega_0 \sin \alpha(t)} + \\ &+ \frac{\cos \epsilon \cos i_0 - \sin i_0 (\sin \Omega_0 \cos \alpha(t) \tan \delta(t) + \sin \epsilon \cos \Omega_0)}{\cos \alpha(t) (\cos \epsilon \sin i_0 \cos \Omega_0 + \sin \epsilon \cos i_0) + \sin i_0 \sin \Omega_0 \sin \alpha(t)}. \end{aligned} \quad (\text{A6.22})$$

This quantity depends on the geometry of the orbit (i_0 , Ω_0) and on the pole position, but it is independent of Ω .

The link between $\frac{d\alpha}{dt}$ and $\frac{d\delta}{dt}$ is a consequence of the circular path of the pole. For $t = 0$ (i.e., at J2000.0) we obtain

$$\hat{\alpha}(0) = 1.77694\hat{\delta}(0). \quad (\text{A6.23})$$

From Equation (A6.18) we get

$$\begin{aligned} \cos \delta \cos \alpha &= u_{V,x1} \cos \beta_V \cos \lambda_V \\ &+ u_{V,y2} \cos \beta_V \sin \lambda_V + u_{V,z1} \sin \beta_V \end{aligned} \quad (\text{A6.24})$$

$$\begin{aligned} \cos \delta \sin \alpha &= \cos \epsilon \\ &(u_{V,x2} \cos \beta_V \cos \lambda_V \\ &+ u_{V,y2} \cos \beta_V \sin \lambda_V + u_{V,z2} \sin \beta_V) \\ &- \sin \epsilon (u_{V,x3} \cos \beta_V \cos \lambda_V \\ &+ u_{V,y3} \cos \beta_V \sin \lambda_V + u_{V,z3} \sin \beta_V) \end{aligned} \quad (\text{A6.25})$$

$$\begin{aligned} \sin \delta &= \cos \beta_V \cos \lambda_V (u_{V,x2} \sin \epsilon \\ &+ u_{V,x3} \cos \epsilon) \\ &+ \cos \beta_V \sin \lambda_V (u_{V,y2} \sin \epsilon \\ &+ u_{V,y3} \cos \epsilon) + \sin \beta_V (u_{V,z2} \sin \epsilon + u_{V,z3} \cos \epsilon). \end{aligned} \quad (\text{A6.26})$$

The functions $\alpha(t)$, $\delta(t)$, obtained by solving Equations (A6.24)–(A6.26), can be expanded to first order in Taylor series around the pole position. We get (units are radians)

$$\begin{aligned} \alpha(t) &\approx -1.52263 + 2.11423\delta\beta_V(t) \\ &- 0.0672264(\delta\lambda_V(t) + \Omega t) \end{aligned} \quad (\text{A6.27})$$

$$\begin{aligned} \delta(t) &\approx 1.17216 - 0.568672\delta\beta_V(t) \\ &- 0.0378326(\delta\lambda_V(t) + \Omega t). \end{aligned} \quad (\text{A6.28})$$

Finally, we obtain the components of the initial velocity of the pole. The evolution of Venus's orbital elements due to

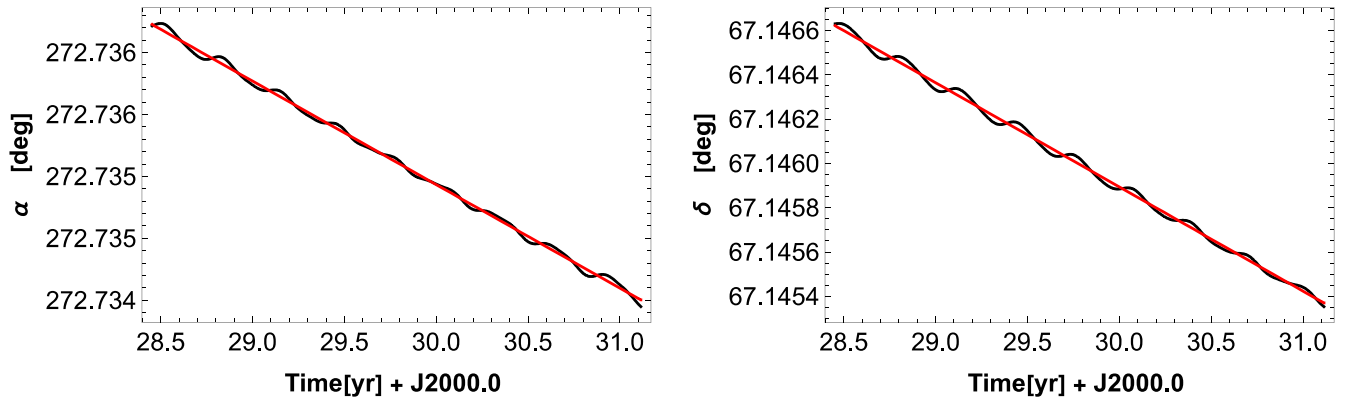


Figure A1. Comparison of time evolution of right ascension (α ; left panel) and declination (δ ; right panel) of the pole when accounting for nutations (black curve) and when nutations are neglected (red curve) over the VERITAS mission time span.

planetary effects is already included in our setup, so here we will consider the solar precession rate only.

From Equations (A6.27) and (A6.28) one can see that while the nutations in latitude (whose amplitudes are smaller than $0''.1$; Cottenreau & Souchay 2009) when projected in α and δ coordinates are about unchanged (rescaling factors are $+2.11$ and -0.57 , respectively), the nutations in longitude (the largest ones, smaller than $3''$ in amplitude) are strongly reduced (rescaled by factors 0.07 and 0.04).

The resulting short-term oscillations, projected on α and δ coordinates, have amplitudes of about $0''.2$ (α) and $0''.1$ (δ), compared with 40 times larger displacements of $8''$ (α) and $4''$ (δ) due to pure precession (see Figure A1).

For this reason, the short-term oscillations do not affect the estimation of $\dot{\alpha}$ and $\dot{\delta}$.

Therefore, neglecting the nutations, the relations between $\dot{\alpha}$, $\dot{\delta}$ and Ω are as follows:

$$\dot{\alpha}(0) \approx -0.0672264\Omega = -4.62097 \times 10^{-13} [\text{rad/s}] \quad (\text{A6.29})$$

$$\dot{\delta}(0) \approx -0.0378326\Omega = -2.60051 \times 10^{-13} [\text{rad/s}]. \quad (\text{A6.30})$$

Finally, the ratio $\dot{\alpha}(0)/\dot{\delta}(0)$ (Equation (A6.22)) can be used as an a priori constraint between the two quantities.

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