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Reanalysis of NOAA H2 observations: implications for the H2 budget / Paulot, F., Pétron, G., Crotwell, A.M., Bertagni, M.B.. - In: ATMOSPHERIC CHEMISTRY AND PHYSICS. - ISSN 1680-7316. - 24:7(2024), pp. 4217-4229. [10.5194/acp-24-4217-2024]

Availability:

This version is available at: 11583/2991481 since: 2024-08-04T10:48:14Z

Publisher:

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Published

DOI:10.5194/acp-24-4217-2024

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Reanalysis of NOAA H₂ observations: implications for the H₂ budget

Fabien Paulot¹, Gabrielle Pétron^{2,3}, Andrew M. Crotwell^{2,3}, and Matteo B. Bertagni^{4,5}

¹Geophysical Fluid Dynamics Laboratory, National Oceanic and Atmospheric Administration, Princeton, NJ, USA

²Cooperative Institute for Research in Environmental Sciences, University of Colorado Boulder, Boulder, CO, USA

³Global Monitoring Laboratory, National Oceanic and Atmospheric Administration, Boulder, CO, USA

⁴High Meadow Environmental Institute, Princeton University, Princeton, NJ, USA

⁵Department of Civil and Environmental Engineering, Princeton University, Princeton, NJ, USA

Correspondence: Fabien Paulot (fabien.paulot@noaa.gov)

Received: 12 July 2023 – Discussion started: 17 July 2023

Revised: 12 January 2024 – Accepted: 9 February 2024 – Published: 9 April 2024

Abstract. Hydrogen (H₂) is a promising low-carbon alternative to fossil fuels for many applications. However, significant gaps in our understanding of the atmospheric H₂ budget limit our ability to predict the impacts of greater H₂ usage. Here we use NOAA H₂ dry air mole fraction observations from air samples collected from ground-based and ship platforms during 2010–2019 to evaluate the representation of H₂ in the NOAA GFDL-AM4.1 atmospheric chemistry-climate model. We find that the base model configuration captures the observed interhemispheric gradient well but underestimates the surface concentration of H₂ by about 10 ppb. Additionally, the model fails to reproduce the 1–2 ppb yr⁻¹ mean increase in surface H₂ observed at background stations. We show that the cause is most likely an underestimation of current anthropogenic emissions, including potential leakages from H₂-producing facilities. We also show that changes in soil moisture, soil temperature, and snow cover have most likely caused an increase in the magnitude of the soil sink, the most important removal mechanism for atmospheric H₂, especially in the Northern Hemisphere. However, there remains uncertainty due to fundamental gaps in our understanding of H₂ soil removal, such as the minimum moisture required for H₂ soil uptake, for which we performed extensive sensitivity analyses. Finally, we show that the observed meridional gradient of the H₂ mixing ratio and its seasonality can provide important constraints to test and refine parameterizations of the H₂ soil sink.

1 Introduction

Increased hydrogen (H₂) usage has been proposed as a strategy to reduce the carbon intensity of many sectors of the economy that are difficult to electrify (Hydrogen Council, 2017; da Silva Veras et al., 2017; Staffell et al., 2019; Abe et al., 2019; Dawood et al., 2020). The climate benefits of greater H₂ usage depend primarily on the H₂ production pathway. Current H₂ production is dominated by steam reforming of methane (CH₄) in natural gas (Holladay et al., 2009; International Energy Agency, 2019), a process that is very carbon intensive (Howarth and Jacobson, 2021). Car-

bon capture can reduce CO₂ emissions associated with H₂ production. However, CH₄ leakage throughout the supply chain could offset much of the expected climate benefits of increased H₂ usage (Howarth and Jacobson, 2021; Ocko and Hamburg, 2022; Bertagni et al., 2022; Hauglustaine et al., 2022). Alternative production pathways such as renewable-based electrolytic H₂ can provide large and rapid reductions in radiative forcing (Hauglustaine et al., 2022), and considerable investments have been devoted to reducing their cost (International Energy Agency, 2022). Furthermore, evidence of significant concentrations of H₂ in surface and subsurface

natural gases (Zgonnik, 2020; Milkov, 2022; Lefeuvre et al., 2021) have spurred interest in the potential of naturally occurring H₂ as a new primary energy source (Prinzhofer et al., 2018; Lapi et al., 2022).

H₂ photooxidation in the atmosphere also tends to increase CH₄, O₃, and stratospheric water vapor, which results in indirect radiative forcing (Derwent et al., 2001; Paulot et al., 2021). Sand et al. (2023) recently calculated that H₂ has a global warming potential of $\simeq 11.6 \pm 2.8$ and 37.3 ± 15.1 for a 100- and 20-year time horizon, respectively.

Significant uncertainties regarding the overall budget of H₂ remain. H₂ sources include both emissions and photochemical production from the oxidation of volatile organic compounds (VOCs). Estimates for the overall source of atmospheric H₂ range from $\simeq 70$ to 110 Tg yr^{-1} , a large spread primarily associated with the magnitude of the H₂ photochemical sources (Ehhalt and Rohrer, 2009). In recent work it has also been argued that current estimates of H₂ sources need to be revised upward to account for geologic H₂ seepage (Zgonnik, 2020). These uncertainties in the nature and magnitude of H₂ sources have proved challenging to reduce, in part because of commensurate uncertainties in H₂ sinks. The atmospheric oxidation of H₂ by OH is well understood but is estimated to account for less than one third of the overall atmospheric sink (Ehhalt and Rohrer, 2009; Paulot et al., 2021). The most important removal pathway is the consumption of H₂ by high-affinity hydrogen-oxidizing bacteria (HA-HOB), a class of bacteria that have been identified in many different soils (Constant et al., 2008; Greening et al., 2015; Bay et al., 2021; Greening and Grinter, 2022). Several parameterizations of the H₂ soil sink have been developed (Ehhalt and Rohrer, 2013; Price et al., 2007; Smith-Downey et al., 2006; Bertagni et al., 2021) that aim at capturing the observed sensitivity of H₂ soil removal to soil temperature, soil moisture, and ecosystem and soil type (Ehhalt and Rohrer, 2009). However, observational constraints on H₂ soil removal remain very limited (Meredith et al., 2016) and this process is challenging to represent in global models (Yashiro et al., 2011; Paulot et al., 2021).

Here, we leverage the recently completed recalibration of H₂ measurements collected by the NOAA Global Monitoring Laboratory to perform a comprehensive evaluation of the simulation of H₂ in the Geophysical Dynamics Laboratory (GFDL) AM4.1 model (Horowitz et al., 2020; Paulot et al., 2021). The NOAA monitoring network provides additional spatial coverage that complements other existing networks (AGAGE (Prinn et al., 2018), CSIRO (Francey et al., 2003)) and offers a unique opportunity to evaluate the skill of the model in capturing changes in H₂ atmospheric concentration since 2010. This period is especially important in gaining a quantitative understanding of the present-day H₂ budget, also given that recent H₂ observations at Mace Head (Derwent et al., 2021, 2023) show both an increase in H₂ concentration and in its soil removal rate. The study is organized as follows: we first describe and evaluate the representation of H₂ in the

GFDL-AM4.1 global chemistry-climate model, focusing on changes in H₂ over the 2010–2019 period. We then assess the sensitivity of the H₂ simulations to uncertainties in the H₂ budget focusing on the representation of anthropogenic H₂ emissions and soil removal.

2 Methods

2.1 Observations

The NOAA Global Monitoring Laboratory (GML) provides long-term monitoring of long-lived greenhouse gases and other trace species. The NOAA GML Global Cooperative Air Sampling Network is a partnership between GML and many outside organizations and individual volunteers to collect discrete air samples approximately weekly from 60+ globally distributed sites (Global Monitoring Laboratory, 2023). These sites are often situated so as to collect air representative of large regional air masses. Priorities are placed on sites where opportunities exist for local support which can be maintained over long (decadal) timescales. The discrete air samples are collected weekly in pairs of 2 L glass flasks and are returned to GML for measurements of multiple species on central measurement systems thus providing a high level of consistency across the globally distributed network.

GML measurements of H₂ in the discrete air samples began in the late 1980s as an opportunistic measurement associated with the analytical technique then used for measuring atmospheric carbon monoxide (CO). To facilitate these H₂ measurements, NOAA/GML developed an in-house H₂-in-air reference scale based on a few gravimetric standards (the latest iteration named “H₂-X1996”). This reference scale was not stable over time and introduced significant time-dependent measurement errors. GML recently converted part of the historical H₂ measurement records to the H₂ calibration scale recommended by the World Meteorological Organization (WMO/MPI H₂-X2009) maintained by the Max Planck Institute (MPI) in Jena, Germany (Jordan and Steinberg, 2011). Measurements since approximately 2010 have been reprocessed onto the MPI scale to remove the biases inherent in the NOAA X1996 scale (Pétron et al., 2024). NOAA-reprocessed H₂ data since 2010 are consistent to within 1–2 ppbv on an annual basis for the same air measurements with CSIRO and the MPI-BGC (Pétron et al., 2024). However, earlier NOAA data that remain on the obsolete NOAA X1996 scale are known to be biased relative to the later NOAA data and to other monitoring programs.

Here, we only consider ground stations from the NOAA Cooperative Air Sampling Network with at least 96 distinct monthly observations over the 2010–2019 period (80 % coverage, Fig. S1 in the Supplement). Ship-based observations are binned in $4^\circ \times 4^\circ$ regions and we only consider regions with at least 40 observations.

2.2 Model setup

We use the GFDL Atmospheric Chemistry Model AM4.1 (Horowitz et al., 2020). For all configurations, the model is run from 2004 to 2019. Monthly sea surface temperature and sea ice concentration are from Rayner et al. (2003) and Taylor et al. (2000). Horizontal winds are nudged to 6-hourly horizontal winds from the National Center for Environmental Prediction (Kalnay et al., 1996). The model output is sampled at the time and location of the air sampling. To better quantify the drivers of the H₂ distribution and trend, we tag H₂ associated with anthropogenic, marine, soil, and biomass burning direct H₂ emissions and with H₂ produced by the oxidation of VOCs.

2.2.1 BASE simulation

AM4.1 includes a detailed representation of H₂ (Paulot et al., 2021), which is briefly summarized here. This configuration will be referred to as “BASE” (Table 1) hereafter. H₂ sources include both direct emissions from anthropogenic and natural sources as well as photochemical production. Anthropogenic emissions of H₂ ($\approx 13 \text{ Tg yr}^{-1}$ over the 2010–2019 period) are estimated from CO emissions in the Community Emissions Data System (CEDS) v20210421 (O’Rourke et al., 2021) using time-invariant sector-specific H₂:CO emission ratios (Table S1 in the Supplement). The transportation and residential sectors are the largest contributors to anthropogenic H₂ emissions (Fig. S2). Biomass burning emissions ($\approx 8 \text{ Tg yr}^{-1}$) are estimated using the Global Fire Emissions Database (GFED4s, van der Werf et al., 2017) with emission factors from Akagi et al. (2011) and Andreae (2019). Marine (6 Tg yr^{-1}) and terrestrial (3 Tg yr^{-1}) sources of H₂ are prescribed as a monthly climatology and distributed spatially (Fig. S3) based on the soil and marine CO emission patterns in the Precursors of Ozone and their Effects in the Troposphere inventory (Granier et al., 2005). The BASE emission inventory does not include geological sources of H₂.

The production of H₂ associated with CH₂O photolysis is calculated interactively using FAST-JX version 7.1, as described by Li et al. (2016). Formaldehyde sources are dominated by the oxidation of VOCs from anthropogenic (O’Rourke et al., 2021), biomass burning (van der Werf et al., 2017), and natural origins. Biogenic emissions of VOCs are prescribed as a monthly climatology (Granier et al., 2005), except for isoprene and terpenes, of which emissions are calculated interactively using the Model of Emissions of Gases and Aerosols from Nature (Guenther et al., 2012). Surface CH₄ is prescribed as a monthly latitudinal profile from observations up to 2014 (Meinshausen et al., 2017) and from the SSP1-2.6 scenario after 2015 (Meinshausen et al., 2020). We select this scenario as it tracks well the observed global CH₄ surface mixing ratio from the World Meteorological Organization Global Atmospheric Watch greenhouse gases observational network (WMO, 2021). To characterize the contri-

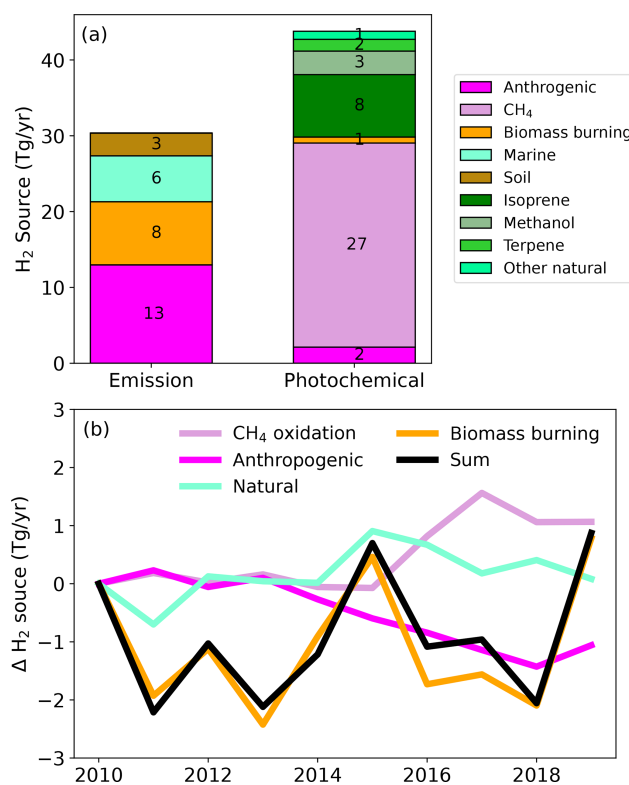


Figure 1. Global source of H₂ (a) and changes in the magnitude of H₂ sources over the 2010–2019 period (b). For clarity, the green line denotes the combined change in H₂ emissions and photochemical production from natural sources (marine and soil emissions + BVOCs photooxidation).

bution of different VOC emissions to the photochemical production of H₂, we perform a set of sensitivity experiments in which we perturb the emission of a given VOC by 10% and quantify the response of H₂ production. For CH₄ oxidation, we directly track the different oxidation pathways that result in H₂ production. The molar yield of H₂ from CH₄, isoprene, methanol, and terpene are estimated to be 0.38, 0.56, 0.21, and 0.70 mol mol⁻¹, respectively. These yields are broadly similar to estimates derived by Ehhalt and Rohrer (2009) (0.37, 0.54, 0.19, and 0.71, respectively) but are lower than estimates derived from a box model (0.38, 0.83, 0.38, and 0.85, respectively for NO_x = 160 pptv; Grant et al., 2010), which may reflect the impact of wet and dry deposition. In particular, Fig. S4 shows that the simulated yield of H₂ from CH₄ oxidation is lowest in the tropics, where most CH₄ is oxidized, as a greater fraction of CH₂O is oxidized by OH in this region than at high latitudes.

Overall, we find that CH₄ oxidation is the largest photochemical source of H₂ ($\approx 27 \text{ Tg yr}^{-1}$). The oxidation of biogenic VOCs (BVOCs) accounts for the majority of the remaining photochemical source of H₂ ($\approx 14 \text{ Tg yr}^{-1}$) primarily from isoprene (8 Tg yr^{-1}), methanol (3 Tg yr^{-1}), and terpene (1.5 Tg yr^{-1}). The oxidation of VOCs from anthro-

pogenic and biomass burning origin produces $\approx 3 \text{ Tg yr}^{-1}$ of H₂. Our estimates are in good agreement with previous estimates (Ehhalt and Rohrer, 2009): CH₄ ($23 \pm 8 \text{ Tg yr}^{-1}$), isoprene ($9 \pm 6 \text{ Tg yr}^{-1}$), biomass burning and anthropogenic VOCs (3 Tg yr^{-1}). This similarity can be attributed to the similar yield of H₂ from CH₂O (0.4 mol mol^{-1} compared to 0.37; Ehhalt and Rohrer, 2009). More work is needed to better characterize the temperature and pressure sensitivity of CH₂O photolysis quantum yields (Röth and Ehhalt, 2015).

Figure 1a summarizes the simulated sources of H₂ associated with photochemical production and direct emissions in the BASE run. Over the 2010–2019 period, the average global simulated source of H₂ is $74 \pm 1 \text{ Tg yr}^{-1}$, with 60 % from photochemical production. Anthropogenic activities are estimated to account for $\approx 40 \%$ of the overall H₂, primarily from CH₄ oxidation. Note that we assume that 50 % of the photochemical production of H₂ from CH₄ oxidation is anthropogenic based on the detailed bottom-up inventory of CH₄ sources (Saunois et al., 2020). Top-down estimates suggest a higher contribution of anthropogenic sources ($\approx 60 \%$; Saunois et al., 2020), which would further increase the fraction of H₂ associated with anthropogenic activities. Figure 1b shows that the simulated total source of H₂ changes little over the 2010–2019 period. The simulated annual photochemical source of H₂ (excluding non-methane VOCs from biomass burning and anthropogenic origins) is 1.6 Tg yr^{-1} greater during 2017–2019 than during 2010–2012, with 70 % of this increase attributed to CH₄. By contrast, H₂ associated with anthropogenic activities decreases (-1.3 Tg yr^{-1} , Fig. S2a), mostly from transport (-1 Tg yr^{-1}) and industries (-0.4 Tg yr^{-1}). The decrease in H₂ emissions reflects the decline in CO emissions from these sectors. The interannual variability of the overall H₂ source during the 2010–2019 period is dominated by the variability of biomass burning emissions, which can result in interannual changes of $\approx 2 \text{ Tg yr}^{-1}$.

H₂ sinks include chemical oxidation by OH and O(¹D) and soil uptake associated with microbial activity. The deposition velocity of H₂ ($v_d(\text{H}_2)$) over land is calculated following the parameterization of Ehhalt and Rohrer (2013) and depends on temperature, soil moisture (Ehhalt and Rohrer, 2013), and soil carbon (Khdhiri et al., 2015; Paulot et al., 2021). In the BASE configuration we use a monthly climatology of $v_d(\text{H}_2)$ calculated using monthly meteorological and soil outputs from the GFDL Earth System Model ESM4.1 over the 1989–2014 period (Dunne et al., 2020; Paulot et al., 2021). Soil uptake is estimated to account for 71 % of the overall H₂ sink. The overall lifetime of H₂ in the BASE configuration is 2.5 years. The lifetime of H₂ associated with anthropogenic emissions is 6 % shorter due to their geographical distribution.

2.2.2 Sensitivity simulations

In this section, we describe additional model simulations that are designed to explore the impact of uncertainties in the representation of H₂ emissions and deposition on the simulation of atmospheric H₂ (Table 1). We focus on H₂ emissions and deposition as their representations in models are largely derived from limited observational constraints (Derwent et al., 2023; Paulot et al., 2021).

The REVISED configuration focuses on the representation of anthropogenic and natural H₂ emissions. The development of the REVISED emission inventory is guided by the biases of the BASE configuration against H₂ observations (Sect. S1.1 in the Supplement, Ghosh et al., 2015). In particular, we focus on the representation of transportation emissions (Table S1) and emissions associated with industrial H₂ use for refining and ammonia, methanol, and steel production. Further details regarding the treatment of anthropogenic and natural sources in the REVISED emission inventory can be found in the Supplement (Sects. S1.2 and S1.3).

We further consider the impact of a different representation of H₂ soil uptake on the simulation of H₂. Here, we use the parameterization of the soil moisture response of HA-HOB activity recently developed by Bertagni et al. (2021). This parameterization relates the minimum soil moisture required for H₂ uptake by HA-HOB to soil hydrological properties, which facilitates its incorporation into global models. This model also allows us to vary the strength of the diffusion barrier associated with soil litter, which can reduce H₂ transport to active sites (Smith-Downey et al., 2008; Ehhalt and Rohrer, 2009). To quantify possible changes in $v_d(\text{H}_2)$ over the 2010–2019 period, we calculate daily deposition velocity using 3-hourly soil moisture, soil temperature, and snow cover from the NASA Global Land Data Assimilation System (Rodell et al., 2004). We focus on two different configurations. In REVISED_GLDAS, we neglect the litter resistance and assume that HA-HOB activity is inhibited when the soil matrix potential (Ψ_{ws}) is less than the wilting point of plants in semiarid environments ($\Psi_{\text{ws}} = -3000 \text{ kPa}$), as recommended by Bertagni et al. (2021). The required soil moisture for the H₂ uptake is not well known and experimental studies have shown that HA-HOB are present in very arid environments (Jordaan et al., 2020). In REVISED_GLDAS2, we assume a much lower activation threshold for HA-HOB ($\Psi_{\text{ws}} = -10\,000 \text{ kPa}$) and account for the litter barrier. Note that both these configurations use the REVISED emission inventory. More details regarding the calculation of $v_d(\text{H}_2)$ can be found in the Supplement (Sect. S1.4).

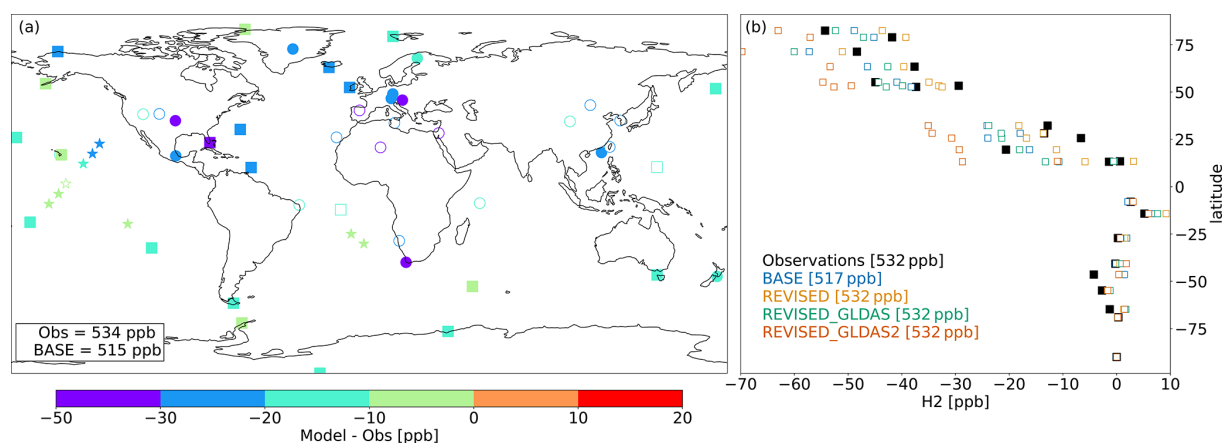


Figure 2. Mean model bias at individual sites for the BASE model configuration (a) over the 2010–2019 period. Filled symbols denote sites where the correlation between observed and simulated H₂ concentrations exceeds 0.5. Squares and stars denote background sites and cruises, respectively. (b) Observed and simulated difference in H₂ at background sites relative to H₂ mole fraction measured at the South Pole Observatory. The average concentrations at background sites are indicated for each configuration in the legend.

Table 1. Model configurations.

	H ₂ anthropogenic emission	H ₂ natural emission	H ₂ soil removal
BASE	Time-invariant H ₂ : CO emission ratio (Table S1)	Ocean + Soil: monthly climatology Biomass burning: GFED4s	Monthly climatology $v_d(\text{H}_2)$ (Paulot et al., 2021)
REVISED	Revised H ₂ : CO emission ratio Emission from industrial H ₂ use (Sect. S1.2 and Table S1)	Ocean: calculated from CO seawater concentration Soil: calculated from N fixation (Sect. S1.3) Biomass burning: same as BASE	Same as BASE
REVISED_GLDAS	Same as REVISED	Same as REVISED	Daily $v_d(\text{H}_2)$ calculated using land reanalysis with soil moisture sensitivity from Bertagni et al. (2021) (Sect. S1.4)
REVISED_GLDAS2	Same as REVISED	Same as REVISED	Same as REVISED_GLDAS with canopy + litter resistance and a lower HA-HOB water-activation threshold (Sect. S1.4)

3 Results and discussion

3.1 BASE model evaluation

3.1.1 Climatology

Figure 2 shows the average model bias against surface observations from NOAA GML. In the BASE configuration, AM4.1 underestimates H₂ at all stations, with greater biases over continental regions (Fig. 2). Correlations exceed 0.5 at more than 90 % of the background sites (squares) but only at 55 % of continental sites. Figure 2b further shows that the concentration at the South Pole is ≈ 50 ppb greater than at the North Pole, which is well captured by the BASE configuration.

To examine differences between the model and the observed seasonality, we first apply the Kmean++ clustering algorithm (Arthur and Vassilvitskii, 2007) to the observed H₂ monthly climatology. Since our focus is on the seasonality of H₂, we transform the monthly climatology of H₂ at each site such that it has a mean of 0 and a standard deviation of 1. Using the within-cluster sum of squares and the silhouette score, we find that the standardized H₂ observations can be well represented using four clusters. Figure 3 shows the seasonality of the standardized H₂ concentration for each cluster (Fig. 3a) and their spatial distribution (Fig. 3b). Sites are found to cluster broadly by latitude based on the seasonality of H₂ with clusters 1, 2, 3, and 4 comprising primarily sites located in the southern mid-to-high latitudes, southern tropics, northern subtropics, and northern mid-to-

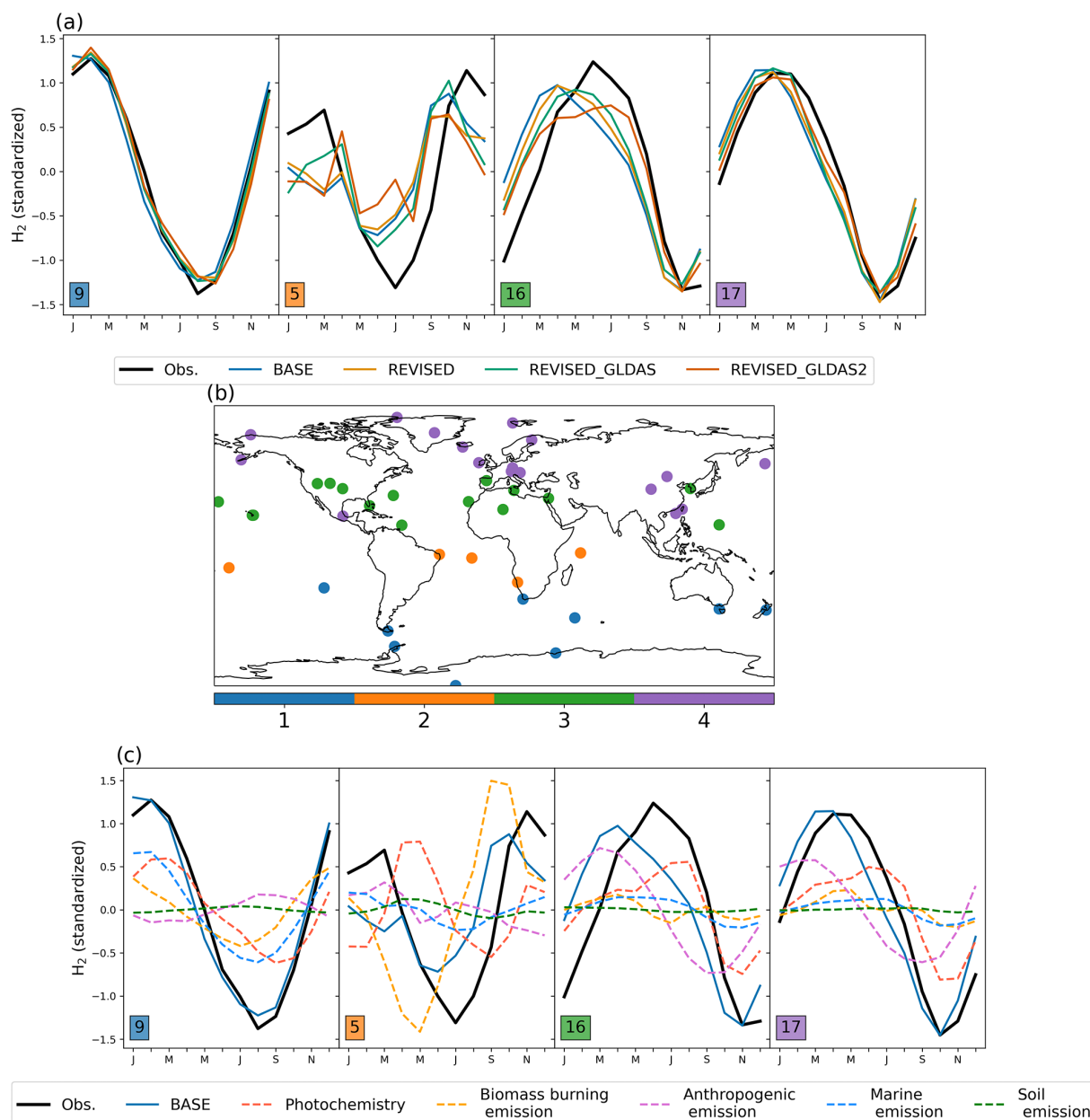


Figure 3. Monthly standardized H₂ concentration for each cluster (a). The number of sites in each cluster is indicated by insets. The sites included in each cluster are shown in panel (b). The variation of source-tagged H₂ tracers in each cluster is shown in panel (c). Source-tagged H₂ tracers are normalized using the standard deviation of simulated H₂.

high latitudes, respectively. The model captures the seasonality of H₂ well in the Southern Hemisphere (cluster 1) but peaks 1–3 months earlier than observations for clusters 2, 3, and 4. Figure 3c shows the contribution of different sources of H₂ to the simulated seasonality of H₂ (inferred from the tagged H₂ tracers). The seasonal bias for cluster 2 is primarily driven by H₂ emitted from biomass burning, which peaks ~2 months earlier than observations. This delay may be associated with greater burning of woody material toward the end of the dry season, emitting more incompletely oxidized

products such as H₂ (van der Werf et al., 2006). Figure 3c also shows that the seasonal bias in clusters 3 and 4 may be associated with H₂ emitted by anthropogenic activities. As we show in Sect. 3.2.2, this seasonal bias may also reflect errors in the removal of H₂.

3.1.2 Time series

Figure 4 shows that H₂ has increased at most sites with an average trend at the background sites of 1.4 ± 0.7 ppb yr⁻¹

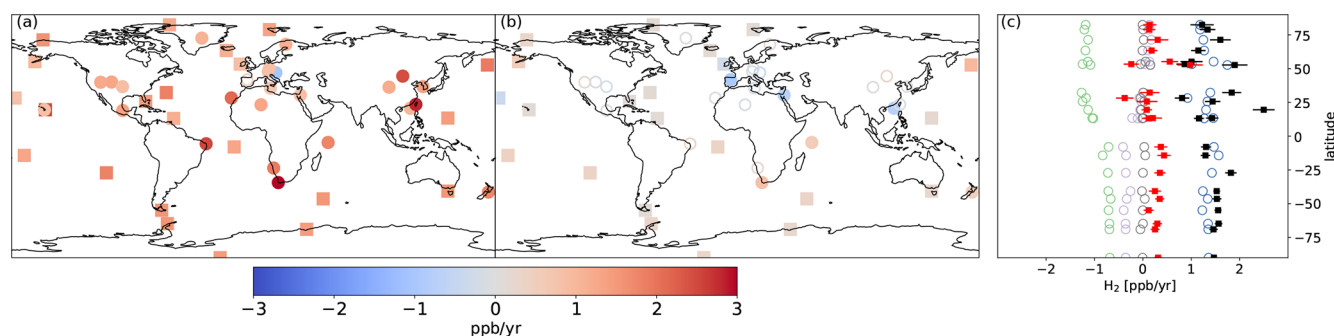


Figure 4. Trend in H₂ concentrations in observations (a) and in the BASE simulation (b) over the 2010–2019 period. (c) Observed (black) and simulated (red) trend in H₂ at background sites (squares) as well as the trend in tagged H₂ tracers associated with anthropogenic sources (green), biomass burning (purple), ocean and soil sources (black), and photochemical production (blue). Filled symbols denote trends that are significantly different from 0 ($p < 0.01$). The error bars show 1 SD for the estimated observed and simulated trends.

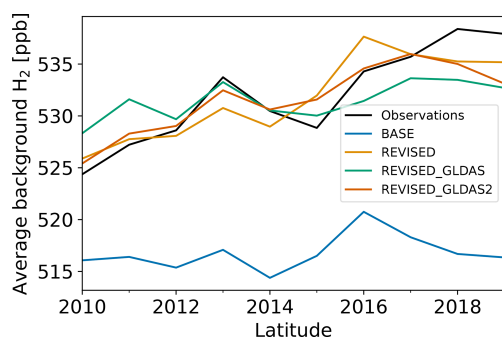


Figure 5. Mean observed and simulated H₂ at background sites (see Fig. 2 for locations).

over the 2010–2019 period with little variability with latitude. Trends are calculated using ordinary least-square regression applied to the de-seasonalized monthly H₂ concentrations. By contrast, the simulated H₂ concentration in the BASE configuration changes little over this time period.

In the Northern Hemisphere, the lack of a trend at background sites in the model (Fig. 4c) reflects the near-cancellation between the increase of photochemically produced H₂ and the decrease in H₂ emitted from anthropogenic sources, consistent with the changes in anthropogenic emissions and the photochemical source of H₂ from the oxidation of CH₄ and biogenic VOCs (Fig. 1). The simulated absolute trend in anthropogenic H₂ is $\simeq 50\%$ lower in the Southern Hemisphere relative to the Northern Hemisphere due to the higher relative areal density of anthropogenic sources in the Northern Hemisphere. By contrast, the change in photochemically produced H₂ exhibits little variability with latitude and matches the observed trend well. The simulated trend also shows little latitudinal variation due to a decrease in H₂ from biomass burning in the Southern Hemisphere.

3.2 Sensitivity simulations

In this section, we explore how uncertainties in the representation of H₂ emissions and deposition contribute to the biases in the BASE model run.

3.2.1 Emissions

Figure 5 shows that the BASE run exhibits a 10–15 ppb negative bias and fails to capture the $\simeq 15$ ppb increase over the 2010–2019 period (Fig. 5). From this bias, we estimate a missing source of H₂ of $\simeq 2$ – 2.5 Tg yr⁻¹ in circa 2010 and 3–4 Tg yr⁻¹ in circa 2019 (Sect. S1.1). Similarly, Derwent et al. (2023) recently reported that a missing H₂ source (5 Tg yr⁻¹ in 2020) was required to explain the observed increase in H₂ concentration at Mace Head and Cape Grim since 2010.

Figures 5 and 6 show that the observed increase in H₂ can be well captured with the REVISED emission inventory. In this inventory, the increase in the missing source of H₂ is explained by a lower decrease in anthropogenic H₂ emissions associated with fossil fuel combustion (0.9 Tg yr⁻¹ lower in 2019 relative to 2010 compared to 1.6 Tg yr⁻¹ in the BASE inventory) and an increase in H₂ emissions associated with H₂ industrial usage (0.3 Tg yr⁻¹). We also increase the H₂ soil source from 3 to 4.5 Tg yr⁻¹ to reduce the model negative bias. This change is well within the large uncertainties in the minor H₂ sources surveyed by Ehhalt and Rohrer (2009). In particular, it is a small fraction of the estimated geological source of H₂ (23 ± 7 Tg yr⁻¹, Zgonnik, 2020), which we do not account for here.

The REVISED emission inventory provides a possible explanation for the observed increase in atmospheric H₂. It highlights the importance of constraining H₂ emissions associated with H₂ industrial use, a sector that is expected to grow rapidly in coming decades.

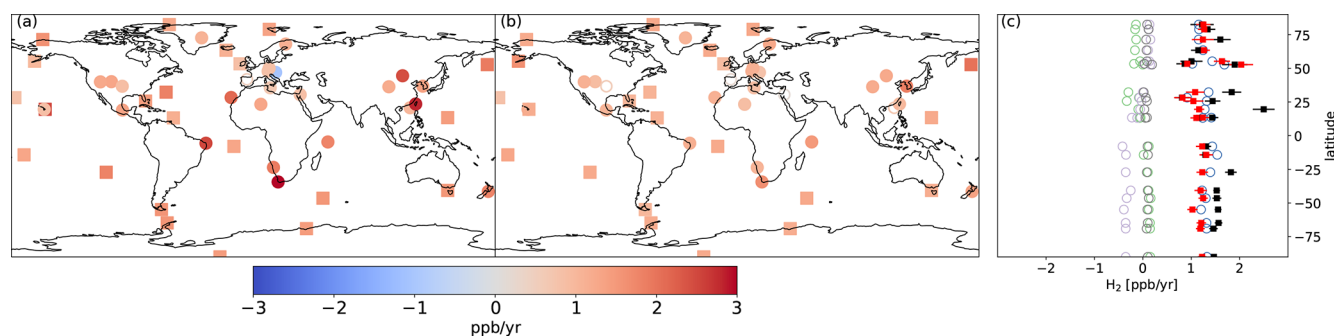


Figure 6. Same as Fig. 4 but for the REVISSED configuration.

3.2.2 Deposition

The BASE and REVISSED experiments assume no interannual variability in $v_d(\text{H}_2)$. However, we have recently shown that climate change may cause an increase in $v_d(\text{H}_2)$ (Paulot et al., 2021). A recent analysis of observations at Mace Head also suggests that $v_d(\text{H}_2)$ has increased in the past few decades (Derwent et al., 2021).

Figure 7 shows that the REVISSED_GLDAS and REVISSED_GLDAS2 $v_d(\text{H}_2)$ exhibit different meridional distributions relative to the BASE configuration with faster removal in the subtropics and northern high latitudes but slower removal in the tropics. This reflects more efficient removal of H₂ in arid regions and slower removal in the tropics. These spatial differences are the largest for the REVISSED_GLDAS2 configuration due to the activation of HA-HOB at a lower soil moisture. Figure 7b further shows that the values of $v_d(\text{H}_2)$ in the REVISSED_GLDAS and in the REVISSED_GLDAS2 configuration both increase over the 2010–2019 period in the northern midlatitudes. This increase reflects drier and warmer conditions in Europe and the Western US as well as parts of Siberia, which result in faster biological uptake rates and promote H₂ diffusivity (Fig. S5). This mechanism may contribute to the reported 1.2 % yr⁻¹ increase in H₂ deposition velocity at Mace Head from 1994 to 2020 (Derwent et al., 2021). Drier conditions in Australia trigger biotic limitations, which results in a large decrease in H₂ deposition velocity in the southern midlatitudes in the REVISSED_GLDAS configuration. By contrast, we find no significant suppression of H₂ uptake in Australia over this time period in the REVISSED_GLDAS2 configuration due a lower threshold for biotic limitations.

Changes to the spatial distribution of $v_d(\text{H}_2)$ and the increase in H₂ removal in the northern midlatitudes (Fig. 7b) in REVISSED_GLDAS result in a larger pole-to-pole difference in surface H₂ (Fig. 2) and a reduction in the simulated trend (Fig. 8) in the northern mid-to-high latitudes. Both of these changes tend to degrade the model performance relative to the REVISSED configuration. By contrast, the REVISSED_GLDAS configuration better captures the timing of

the H₂ maximum in the Northern Hemisphere (clusters 3 and 4, Fig. 3a).

Figure 9 shows the systematic assessment of the sensitivity of $v_d(\text{H}_2)$ to Ψ_{ws} and the strength of the litter barrier. We find that a lower soil moisture threshold for HA-HOB activation (i.e., a lower Ψ_{ws}) favors H₂ removal in the Northern Hemisphere relative to the Southern Hemisphere (Fig. 9a) and results in a larger increase in $v_d(\text{H}_2)$ over the 2010–2019 period (Fig. 9b), especially in the Southern Hemisphere (Fig. 9c). This suggests that a lower Ψ_{ws} would tend to worsen the model performance in the absence of a litter barrier (given the REVISSED emissions). The litter barrier tends to increase the importance of arid regions for H₂ removal. This makes H₂ uptake more susceptible to moisture inhibition, such that a stronger litter barrier tends to result in a lower increase or even a decrease in $v_d(\text{H}_2)$ over the 2010–2019 period. Under all scenarios, the litter barrier tends to increase the gradient in $v_d(\text{H}_2)$ between the Northern Hemisphere and the Southern Hemisphere.

It is notable that no configuration results in a small change in $v_d(\text{H}_2)$ without producing large and increasing gradients between the Northern Hemisphere and Southern Hemisphere. As a result, our model cannot capture the observed trends, meridional gradient, and seasonality together given our REVISSED estimate of H₂ emissions. This is illustrated by the REVISSED_GLDAS2 configuration ($\Psi_{\text{ws}} = -10\,000$ kPa, litter_scale = 1), which is found to improve the simulated trend relative to the REVISSED_GLDAS (not shown) and the simulated seasonality relative to the REVISSED configuration (Fig. 3) but results in a large overestimate of the hemispheric gradient (Fig. 2). This highlights the need for a more detailed representation of the factors that modulate $v_d(\text{H}_2)$ (Khdhiri et al., 2015) to help interpret changes in H₂ concentrations.

4 Conclusions

The recently released H₂ dry air mole fraction measurements from the NOAA Global Cooperative Air Sampling Network expand the spatial coverage of the WMO Global Atmospheric Watch observations. This offers the opportu-

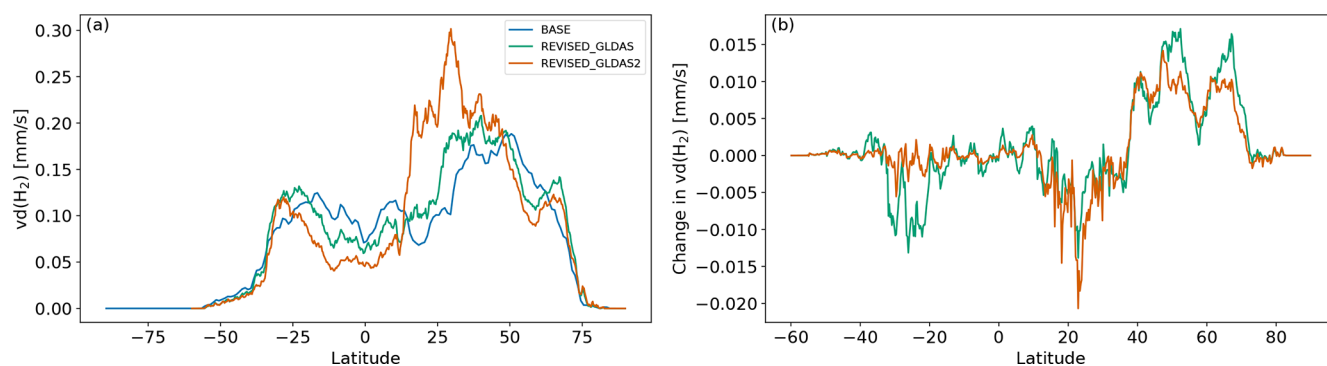


Figure 7. Meridional distribution of $v_d(\text{H}_2)$ in the BASE, REVISED_GLDAS, and REVISED_GLDAS2 simulations (a) and (b) simulated change in $v_d(\text{H}_2)$ between the periods 2017–2019 and 2010–2012.

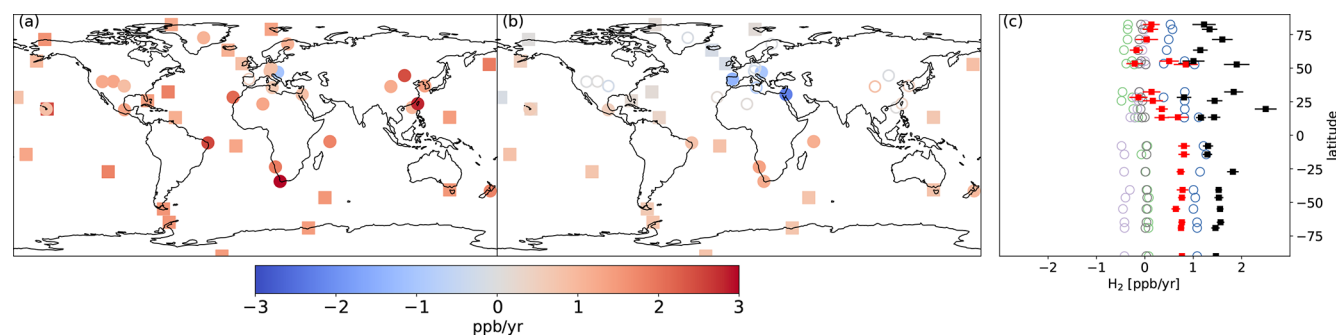


Figure 8. Same as Fig. 4 but for the REVISED_GLDAS configuration.

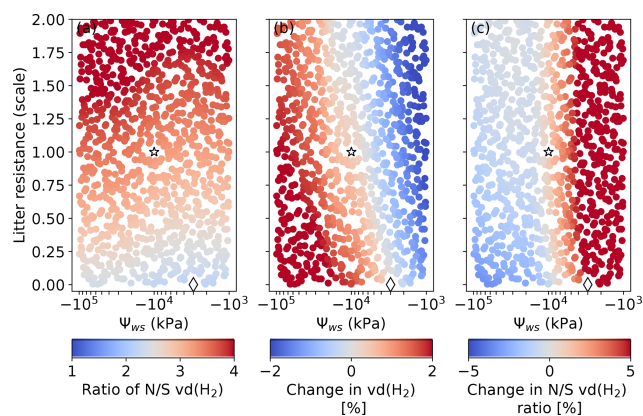


Figure 9. Simulated sensitivity of $v_d(\text{H}_2)$ to Ψ_{ws} and the strength of the litter diffusive barrier. Panel (a) shows the response of the north/south ratio of $v_d(\text{H}_2)$. Panel (b) shows the change in $v_d(\text{H}_2)$ between 2010–2012 and 2017–2019. Panel (c) shows the change in the north/south ratio of $v_d(\text{H}_2)$ between 2010–2012 and 2017–2019. The REVISED_GLDAS configuration uses $\Psi_{ws} = -3000$ kPa and no litter resistance (diamond). The REVISED_GLDAS2 uses $\Psi_{ws} = -10000$ kPa and the default (scale = 1) litter resistance (star). The litter scale reflects the perturbation to the default litter resistance (see Sect. S1.4).

nity to assess the representation of the H₂ atmospheric budget in the state-of-the-art GFDL-AM4.1 global atmospheric chemistry-climate model. Observations show that H₂ has increased on average by 1–2 ppb yr⁻¹ over the 2010–2019 period. This change can be explained by the increase in photochemically produced H₂ (mostly from CH₄), provided direct anthropogenic H₂ emissions have remained stable during this time period. We hypothesize that this stability reflects the compensation between declining emissions associated with fossil fuel combustion (mostly from the transport sector) and increasing emissions associated with H₂-producing facilities (for refining and ammonia, methanol, and steel production). This is notable since H₂ release from H₂ production facilities is poorly understood yet important for assessing the climate benefits of H₂ (Hauglustaine et al., 2022; Bertagni et al., 2022).

We show that the observed trend, seasonality, and meridional gradient of H₂ provide complementary constraints on the global H₂ biogeochemical cycle. We find that our model fails to capture all three constraints together, which most likely reflects fundamental gaps in our representation of the soil removal of H₂ by microorganisms (HA-HOB). Such uncertainties are important, as an increase in $v_d(\text{H}_2)$ would require a commensurate increase in H₂ sources to explain the observed change in H₂ concentration.

This study highlights the need for coordinated field and laboratory data collection efforts to help improve models of the distribution and activity of HA-HOB in global models (American Academy of Microbiology, 2023). Such work is critical for quantifying the response of atmospheric H₂ to increasing anthropogenic H₂ usage as well as hydrological changes associated with climate change (Jansson and Hofmockel, 2019; Huang et al., 2015); however, it is hindered by the lack of sensors that offer higher time resolution and maintain good sensitivity and stable response.

Code and data availability. The code for the GFDL ESM4.1 model is available at <https://doi.org/10.5281/zenodo.5347705> (Robinson, 2021). NOAA Global Cooperative Network Flask Air H₂ (Pétron et al., 2023) can be downloaded at <https://doi.org/10.15138/WP0W-EZ08>.

Supplement. The supplement related to this article is available online at: <https://doi.org/10.5194/acp-24-4217-2024-supplement>.

Author contributions. FP designed the research and developed and analyzed the model simulations. GP and AC collected and processed H₂ observations from the NOAA network and provided guidance regarding their interpretation. MB developed the soil moisture parameterization of HA-HOB used in the REVISED_GLDAS and REVISED_GLDAS2 configurations. All authors contributed to the drafting of the manuscript.

Competing interests. The contact author has declared that none of the authors has any competing interests.

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Acknowledgements. We thank Vaishali Naik for her help with generating model-ready H₂ emissions. We thank Larry Horowitz, Vaishali Naik, Amilcare Porporato, and Xinning Zhang for their helpful comments on the paper.

Financial support. This research has been supported in part by NOAA cooperative agreements NA17OAR4320101 and NA22OAR4320151 and by the US Department of Energy, Office

of Energy Efficiency and Renewable Energy (EERE), specifically the Hydrogen and Fuel Cell Technologies Office.

Review statement. This paper was edited by Manvendra Krishna Dubey and reviewed by two anonymous referees.

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