

Modeling on the drought stress impact on the summertime biogenic isoprene emissions in South Korea

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Abstract. Biogenic isoprene emissions play an important role in air quality so it is important to quantify their response to extreme events such as drought. This study aims at constraining drought stress on biogenic isoprene emissions in South Korea using satellite formaldehyde (HCHO) column, the key product of isoprene oxidation, and a chemistry transport model (GEOS-Chem) with Model of Emissions of Gases and Aerosols from Nature version 2.1 (MEGAN2.1). It was found that the HCHO column from the Ozone Monitoring Instrument (OMI) increased by 5.4 % under the drought condition compared to the normal condition, but GEOS-Chem simulated a 20.23 % increase indicating an overestimation of isoprene emissions under drought. We implemented two existing drought stress algorithms in MEGAN2.1 and found they were not effective to reduce HCHO column biases in South Korea because those algorithms were proposed and developed for the Southeast United States (SE US). To improve this, we applied an iterative finite difference mass balance (IFDMB) method to estimate isoprene emissions using the OMI HCHO column. With this method, isoprene emissions were reduced by 60 % under the drought condition compared to those simulated by the standard MEGAN2.1 implemented in GEOS-Chem. The increase of HCHO column under the drought condition compared to the normal condition was also reduced to 10.71 %, which is comparable to that in the satellite retrievals. Based on isoprene emission difference between MEGAN2.1 and IFDMB, we developed the empirical equations to adjust isoprene emissions in South Korea that also improved model prediction of the secondary pollutant such as ozone.

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1 Introduction

In the troposphere, biogenic non-methane volatile organic compounds (BVOCs) are important to regional air quality because they are precursors to tropospheric ozone and secondary organic aerosols (Atkinson, 2000; Holm and Balmes, 2022; Pacifico et al., 2009). BVOCs are emitted from terrestrial vegetations and 50–70 % of global BVOCs emissions are isoprene emissions (Pacifico et al., 2009; Sindelarova et al., 2014). Isoprene emissions depend on not only physiological factors such as plant functional type, leaf area index, and leaf age, but also meteorological factors such as temperature, radiation, and soil moisture, which affect plant's physiology such as stomatal conductance, isoprene synthase activity, and carbon substrate supply (Ferracci

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et al., 2020; Guenther et al., 2012; Guenther et al., 2006; Potosnak et al., 2014; Seco et al., 2022). Given this, it is of great
30 importance in the air quality study to quantify such climatic or meteorological impacts on isoprene emissions.

Drought is one of the most important climatic or meteorological extreme events that can modulate isoprene emissions by high
temperature and low soil moisture (Limousin et al., 2010; Zhou et al., 2014; Rissanen et al., 2022; Huang et al., 2015; Seco et
al., 2015). Although the response of isoprene emissions is non-monotonic to the drought severity, isoprene emissions tend to
increase under drought condition in general (Ferracci et al., 2020; Funk et al., 2005; Pegoraro et al., 2004; Potosnak et al.,
35 2014) while the extent of the increase is challenging to quantify because the number of in-situ isoprene measurements
worldwide is spatiotemporally scarce. Taking advantage of wide spatiotemporal coverage of satellite retrievals, some studies
(Wang et al., 2022b; Wasti and Wang, 2022) have used the tropospheric formaldehyde (HCHO) column retrievals from the
satellite to estimate isoprene emissions response to drought. It is well known that HCHO could be used as a proxy for isoprene
emissions because HCHO is produced fast with high yield from isoprene oxidation (Wolfe et al., 2016; Palmer et al., 2003;
40 Sprengnether et al., 2002). Recent studies (Li et al., 2022; Naimark et al., 2021) showed that the tropospheric HCHO column
from the Ozone Monitoring Instrument (OMI) on the Aura satellite increased by 6.5–22 % in the southeastern United States
(US) region during the summertime drought, which is indicative of the increase of isoprene emissions during drought.

The Model of Emissions of Gases and Aerosols from Nature (MEGAN) (Guenther et al., 2006; Guenther et al., 2012) is widely
used to estimate isoprene emissions as a function of physiological and meteorological factors. However, isoprene emissions
45 simulated by MEGAN showed a large overestimation under the drought condition, so some efforts have been carried out in
various ways to reduce such bias (Wang et al., 2022a; Wang et al., 2021). In Jiang et al. (2018), for example, a drought stress
algorithm for MEGAN isoprene emission was proposed based on observational constraints from isoprene flux measurements
at the Missouri Ozarks AmeriFlux (MOFLUX) site in the southeastern US (SE US) during two summertime drought condition
(summers of 2011 and 2012). Wang et al. (2022b), on the other hand, used the OMI HCHO column as top-down constraints
50 to derive a drought stress algorithm for MEGAN isoprene emissions in order to achieve a wider spatiotemporal coverage. They
compared the OMI HCHO column with the HCHO column simulated by a chemical transport model (GEOS-Chem) with
MEGAN isoprene emission in the SE US region for a multiple-year period (2005–2017 summers) and showed that the model
HCHO column had larger biases under the drought condition compared to the normal/wet condition. To reduce the model
HCHO biases under the drought condition, they derived a drought stress algorithm for MEGAN isoprene emission which could
55 minimize the HCHO biases in the SE US region. This algorithm reduced isoprene emission by 8.6–20.7 % in the summertime
drought condition in the SE US.

In South Korea, it is also known that isoprene emissions play an important role on air quality (Kim et al., 2018; Kim et al.,
2013; Lee and Park, 2022). Observational evidence of the impact of drought stress on isoprene emissions (Wasti and Wang,
2022) showed that the OMI HCHO column increased by 2.97 % in the mild drought and by 8.02 % in the extreme drought in

60 the summertime in the South Korean region, indicating the increase of isoprene emissions under the drought condition. However, an effort to evaluate MEGAN isoprene emissions in South Korea has not been made yet despite the wide use of MEGAN in air quality simulations for South Korea (Jang et al., 2020; Lee and Park, 2022; Lee et al., 2014; Yu et al., 2023). Following this need, the present study aimed to constrain isoprene emissions under the summertime (June-July-August, JJA hereafter) drought condition in the South Korean region by using the chemical transport model (GEOS-Chem) with MEGAN
65 and satellite HCHO columns. We first investigated the HCHO column biases of GEOS-Chem under the drought condition. After we verified that the current two drought stress algorithms for MEGAN isoprene emissions were not effective to reduce the HCHO column biases in South Korea, we conducted a top-down inversion of isoprene emissions in this region and estimated the drought impact on isoprene emissions based on the inversion results.

2. Data and Method

70 2.1. Drought index

We adopted a gridded drought index, the evapotranspiration deficit index (DEDI), to detect the drought in South Korea region. The DEDI is derived from actual evapotranspiration (AET) and potential evapotranspiration (PET) datasets (Zhang et al., 2023), which are provided by the European Centre for Medium-Rand Weather Forecast (ECMWF) Reanalysis version 5 (ERA5) (Hersbach et al., 2020). Because the DEDI considers a balance between the atmospheric evaporation demand and the
75 actual land water evaporated from soil, water surfaces and vegetation, the DEDI can connect the climate system with terrestrial ecosystems (Zhang et al., 2023). The horizontal resolution of the DEDI data was $0.25^{\circ} \times 0.25^{\circ}$ and it was regridded to $0.25^{\circ} \times 0.3125^{\circ}$ to match with the horizontal resolution used in the following model simulations. Consistent with the previous studies (Wang et al., 2022b), we focused on the weekly drought by averaging daily DEDI values in all grid points. To decide the drought thresholds of the DEDI data in South Korean region, we followed the percentile category approach. As described in
80 the previous studies (Zhang et al., 2023; Zhang et al., 2022), the 30 % percentile value (-0.49) of the climatological DEDI (1981 – 2010; 30 years) was obtained to detect drought periods ($DEDI \leq -0.49$). **Since DEDI data is available at fine spatial ($0.25^{\circ} \times 0.25^{\circ}$) and temporal (daily) resolutions than other drought indices, we chose DEDI over other indices to help increase data sizes in the limited study period and domain.** Figure S1 shows the weekly DEDI values in the recent seven summers (2016 – 2022 JJA) and compared them with the Standardized Precipitation Index 1 (SPI1) (McKee et al., 1993) which is based on the
85 1-month accumulated precipitation measurements from the in-situ sites in South Korea. The daily SPI1, provided by the Korea Meteorological Administration (KMA), was averaged to derive the weekly SPI1, and the DEDI value at the nearest point to each SPI1 in-situ site was used here for comparison with the SPI1. As described in Zhang et al. (2023), DEDI had consistent drought conditions with the SPI1 (Fig. S1), indicating that DEDI can represent the observed drought conditions. Both drought indices showed that 2016 JJA, 2017 JJA, and 2018 JJA had strong and persistent drought conditions compared to other
90 summers, so we focused on these three summers (2016 – 2018 JJA) in the following analyses and model simulations. **For each**

grid point, therefore, the normal ($DEDI > -0.49$) and drought ($DEDI \leq -0.49$) conditions were defined based on a weekly basis using DEDI during three summers (2016 – 2018).

2.2 In-situ observations of Ozone and PM_{2.5} and the OMI HCHO product

We used the observed in-situ surface ozone (O_3) and particulate matter with a diameter less than or equal to $2.5\mu m$ (PM_{2.5}) datasets in South Korea for 2016 – 2018 JJA period. The hourly O_3 (PM_{2.5}) concentrations at all available sites were used in this study (Fig. 1) to derive daytime mean (7am – 6pm) O_3 (PM_{2.5}) concentrations in each day. We focused on daytime because most of the biogenic isoprene emissions occur in daytime and BVOCs react with nitrogen oxides (NO_x) under the presence of solar radiation to produce secondary pollutants, especially O_3 (Atkinson, 2000). The weekly daytime-mean O_3 (PM_{2.5}) concentration at each site was derived by averaging daily daytime mean O_3 (PM_{2.5}) concentrations.

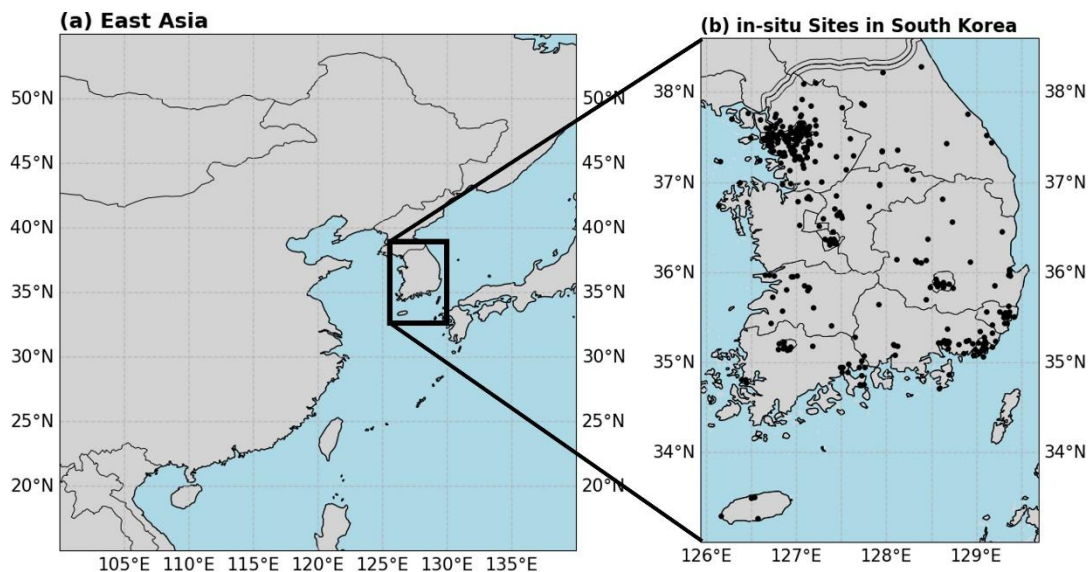


Figure 1: (a) The location for South Korea in East Asia and (b) the locations for in-situ air pollutant measurements in South Korea.

The Ozone Monitoring Instrument (OMI) v003 Level 3 tropospheric formaldehyde (HCHO) column density data (OMHCHOd; Chance, 2019) were used in this study. All pixels with bad formaldehyde retrievals, high cloud fractions ($>30\%$), high solar zenith angle ($>70^\circ$), and pixels affected by OMI's row anomaly in Level 2 HCHO column data (OMHCHO) have been already filtered out in OMHCHOd data (Chance, 2019). The horizontal resolution was $0.1^\circ \times 0.1^\circ$, and it was regridded to $0.25^\circ \times 0.3125^\circ$ to match with the horizontal resolution used in the model simulations. Zhu et al. (2020) verified the OMI HCHO column retrievals with the HCHO aircraft observation in South Korea region during the Korea–United States Air Quality

(KORUS-AQ) campaign from the May to early June 2016. They found that the OMI HCHO column retrievals were biased low by about 22 % relative to the aircraft observation. To correct this underestimation, we applied a constant factor of 1.28 ($1 / (1-0.22)$) to the OMHCHO data as in the previous studies (Shen et al., 2019; Wang et al., 2022b). We also compared this method to a different bias correction method suggested by Müller et al. (2024) and the results were shown in Figure S2. The two corrected HCHO columns were highly correlated with a regression slope close to one, except for a few HCHO points under normal condition. The corrected HCHO columns used in this study were 7-12% higher compared to those from Müller et al. (2024) under normal and drought conditions, respectively.

2.3 GEOS-Chem model simulations

We used the atmospheric composition simulations of the nested-grid GEOS-Chem chemical transport model (version 12-7-2) over East Asia (100°E – 140°E , 15°N – 55°N ; Fig. 1) with a horizontal resolution of $0.25^{\circ} \times 0.3125^{\circ}$ (Bey et al., 2001; Chen et al., 2009; Wang et al., 2004) to obtain the modeled isoprene emissions and HCHO columns in 2016 – 2018 JJA period. The model was driven by GEOS-forward processing (GEOS-FP) meteorological data from NASA’s Global Modeling and Assimilation Office (GMAO). The chemical boundary conditions for the nested domain were obtained from the GEOS-Chem global simulations with a horizontal resolution of $2^{\circ} \times 2.5^{\circ}$ every three hours. The GEOS-Chem Tropchem mechanism was used for the gas-phase chemistry mechanisms in GEOS-Chem (Fisher et al., 2016; Mao et al., 2013; Marais et al., 2016) and gas-aerosol phase partitioning of the nitric acid and ammonia was calculated by ISORROPIA II (Fountoukis and Nenes, 2007). For anthropogenic emissions in East Asia region, we used the Korea-United States Air version 5 (KORUSv5) emission inventory (Woo et al., 2020), which is one of the most recent anthropogenic emission inventories in East Asia. The Global Fire Emissions Database 4 inventory (GFED4) (Van Der Werf et al., 2010) was used for biomass burning, and biogenic emissions such as isoprene emissions were simulated by MEGAN version 2.1 (MEGAN2.1) (Guenther et al., 2012). The details of biogenic isoprene emissions in MEGAN2.1 are described in the following section.

2.4 Isoprene emissions in MEGAN2.1 and two drought stress algorithms used in this study

In GEOS-Chem, biogenic isoprene emission factor ($\gamma_{2.1}$) in MEGAN2.1 is represented as below (Guenther et al., 2006; Guenther et al., 2012):

$$\gamma_{2.1} = C_{\text{FAC}}\gamma_{\text{PAR}}\gamma_{\text{T}}\gamma_{\text{AGE}}\gamma_{\text{LAI}}\gamma_{\text{CO2}}\gamma_{\text{SM}} = \gamma_0\gamma_{\text{SM}} \quad (1)$$

where C_{FAC} is a canopy environment coefficient, γ_{PAR} is a factor for light, γ_{T} is a factor for temperature, γ_{AGE} is a factor for leaf age, γ_{LAI} is a factor for leaf area index (LAI), γ_{CO2} is a factor for CO_2 inhibition, and γ_{SM} is a factor for soil moisture. Here, the factors except for γ_{SM} can be referred to as γ_0 to represent non-drought factors. γ_{SM} is set to one by default in the standard GEOS-Chem because there is no available soil moisture database. Therefore, this limitation can lead to overestimation of isoprene emissions and the following HCHO column under the drought condition (Wang et al., 2022b).

We implemented two existing drought stress algorithms proposed for the better representation of isoprene emissions in MEGAN under drought condition and tested them in the South Korean region. One drought stress algorithm implemented in this study was proposed in Wang et al. (2022b). Wang et al. (2022b) found that GEOS-Chem overestimated HCHO column in the SE US compared to the OMI HCHO column under drought condition and this overestimation was larger in high surface temperature condition (above 300K). The drought stress algorithm was derived to minimize these HCHO column biases between OMI and GEOS-Chem, and the resultant formula for γ_{SM} is shown below:

$$\gamma_{SM} = \begin{cases} 1 & (\beta_t \geq 0.64 \text{ or } T \leq 300K): \text{Non-drought} \\ 380.10e^{-0.02T} & (\beta_t < 0.64 \text{ or } T > 300K): \text{Drought} \end{cases} \quad (2)$$

where T is surface temperature and β_t is soil moisture stress. To get β_t value at each grid in GEOS-Chem, an ecophysiology module developed by Lam et al. (2023) was adopted in GEOS-Chem. The basic formulations in the ecophysiology module are based on the Joint UK Land Environmental Simulator (JULES) (Best et al., 2011; Clark et al., 2011), and this ecophysiology module can calculate soil moisture stress (β_t) using gridded soil parameter data from Hadley Centre Global Environment Model version 2–Earth System Model (HadGEM2-ES). This soil moisture stress (β_t) ranges from zero (fully stressed) to one (no stress) and it can be used as a threshold to trigger the drought stress algorithm in MEGAN2.1. In this study, the threshold for the β_t was set to 0.64 which was 60 % percentile of β_t in the South Korean drought condition (Fig. S3), similar to the threshold used for the SE US (0.60) in Wang et al. (2022b).

The other drought stress algorithm implemented in this study was proposed in Jiang et al. (2018). Based on the observed physiological response of vegetations to drought stress, this drought stress algorithm was derived by reducing V_{cmax} , which is a maximum carboxylation rate by photosynthetic RuBisCo enzyme in vegetations (Jiang et al., 2018). The resultant formula for γ_{SM} is shown below:

$$\gamma_{SM} = \begin{cases} 1 & (\beta_t \geq 0.64): \text{Non-drought} \\ V_{cmax}/\alpha & (\beta_t < 0.64, \alpha = 77): \text{Drought} \end{cases} \quad (3)$$

The V_{cmax} value is also available from the ecophysiology module in GEOS-Chem. The α value of 77 was derived to minimize the mean bias of isoprene fluxes between GEOS-Chem and the measurement at the Missouri Ozarks AmeriFlux (MOFLUX) site under the SE US drought condition (Wang et al., 2022b).

It is noteworthy that these two drought stress algorithms were developed or tuned based on observational constraints pertaining to the SE US region. In the next section, we investigated whether these two drought stress algorithms work in the South Korean region or not. Given that the HCHO column is the main constraint of isoprene emissions, the investigation focused on the comparison between the OMI HCHO column and the HCHO column simulated with these two drought stress algorithms. For

convenience purposes, the simulations with drought stress algorithms by Wang et al. (2022b) and by Jiang et al. (2018) were referred to as WD and JD, respectively.

3. HCHO column biases under the drought condition in South Korea

3.1 HCHO column biases in the standard GEOS-Chem

170 Before investigating the two drought stress algorithms, we first compared the HCHO columns from the OMI and from the standard GEOS-Chem under the normal and drought conditions in the South Korean region (Fig. 2). The mean OMI HCHO column in South Korea was 1.11×10^{16} molec. cm^{-2} under the normal condition during the study period (2016 – 2018 JJA) and increased by 5.4 % to 1.17×10^{16} molec. cm^{-2} under the drought condition (Figs. 2a-c). Wasti and Wang (2022) showed comparable increases of the OMI HCHO column under the South Korean drought condition, which were 2.97 % in the mild
175 drought and 8.02 % in the extreme drought. An increase of similar percentages in HCHO columns was also found in other regions under drought condition such as the SE US and Amazon (Li et al., 2022; Morfopoulos et al., 2022; Wang et al., 2022b). It is known that the impact of water stress on isoprene emissions depends on the severity of the drought (Potosnak et al., 2014; Seco et al., 2015; Wasti and Wang, 2022). To examine this impact, the DEDI was separated into five categories following Zhang et al. (2022): Normal/Wet, Incipient drought, Moderate drought, Severe drought, and Extreme drought (Table S1).
180 Based on these five categories, we calculated the observed HCHO columns in each drought category (Fig. S4). While the domain-mean HCHO columns tended to increase as the drought became stronger, the signal is weak and not uniform by location. For example, the OMI HCHO column over the northeastern parts of South Korea (Taebaek Mountains), which showed a decrease under the drought condition (Figs. 2a-c), showed an increase only under the extreme drought category (Fig. S4). This is probably because our study period included only three summers. Given the small sampling size, we chose not to
185 separate drought severity in the following analysis. By comparison, the mean HCHO columns simulated by the standard GEOS-Chem was 1.33×10^{16} molec. cm^{-2} under the normal condition and increased by 20.23 % to 1.59×10^{16} molec. cm^{-2} under the drought condition (Figs. 2d-f). Thus, the model was found to significantly overestimate the increase of the HCHO column under the drought condition compared to that of the OMI HCHO column. This overestimation can be further revealed by the model biases under both normal and drought conditions. The model (relative) biases of the HCHO column were 0.22×10^{16}
190 molec. cm^{-2} (19.82 %) under the normal condition and 0.42×10^{16} molec. cm^{-2} (35.89 %) under the drought condition (Table 1). The model bias increased by 0.20×10^{16} molec. cm^{-2} (16.07 %) under the drought condition, indicating larger HCHO column bias under the drought condition. The spatial correlation between the OMI HCHO column and the standard GEOS-Chem under the normal condition was 0.58 (Fig. S5a), which was statistically significant at a 99% confidence level based on the Student's *t*-test. It indicates that the GEOS-Chem has reasonable performance in terms of spatial distribution of the HCHO column under
195 the normal condition. However, the lower spatial correlation between two (0.32) was found under the drought condition (Fig. S5b), which consistently indicated the worsening performance of GEOS-Chem under the drought condition. The worsening

performance of GEOS-Chem under the drought condition was also found in other regions. In the SE US region (Wang et al., 2022b), for example, GEOS-Chem had a minimal bias of the HCHO column (0.05×10^{16} molec. cm^{-2}) under the normal condition, and the biases tended to increase with the drought severity (0.08×10^{16} molec. $\text{cm}^{-2} - 0.15 \times 10^{16}$ molec. cm^{-2}), resulting in 1.1 – 1.5 times higher increases of the HCHO column under the drought condition compared to that of the OMI HCHO column.

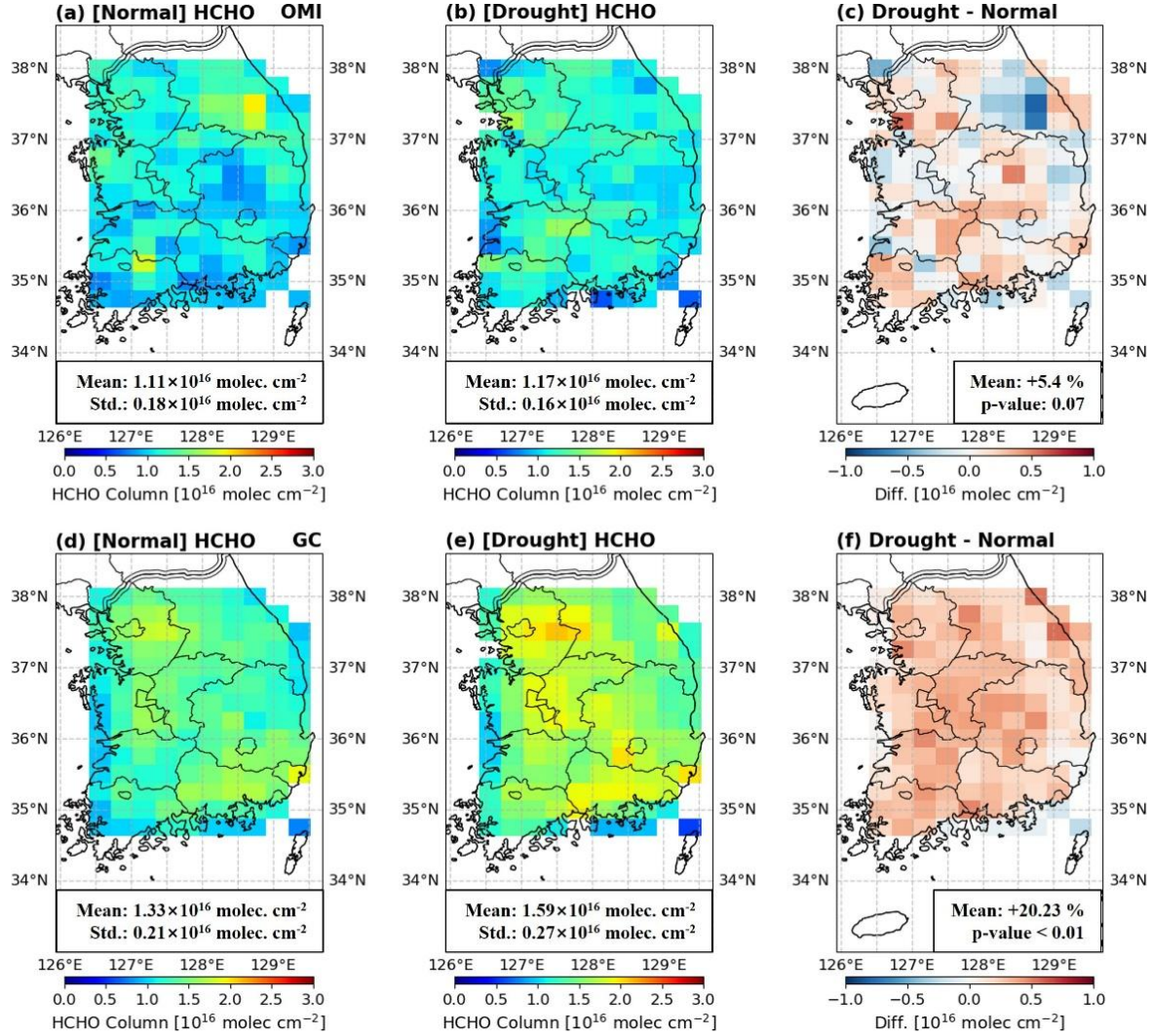


Figure 2: The OMI HCHO column under (a) the normal condition, (b) the drought condition, and (c) the difference (drought – normal). (d-f) Same as a-c but for the standard GEOS-Chem. The mean and standard deviation (Std.) is presented in each panel and p-value for the mean difference was derived from Student's *t*-test.

Table 1: The mean HCHO column bias (relative bias) of GEOS-Chem simulations under the normal condition and drought condition in South Korean region.

Unit: 10^{16} molec. cm^{-2}	Standard GEOS-Chem	WD	JD
Normal	0.22 (19.82 %)	0.18 (16.22 %)	0.13 (11.71 %)
Drought	0.42 (35.89 %)	0.36 (30.77 %)	0.26 (22.22 %)

As described in the aforementioned study (Wang et al., 2022b), the worsening performance of GEOS-Chem under the drought condition suggested an improper representation of the drought-induced change in the model. Given that HCHO is produced by oxidation of VOCs emitted to the atmosphere, we examined how the anthropogenic VOC (AVOC) and BVOC emissions over South Korea changed under the drought condition compared to the normal condition in the model (Fig. 3a). Here, following Lee and Park (2022), AVOC emissions included toluene, xylene, lumped alkanes ($\geq \text{C}_4$), benzene, acetaldehyde, lumped alkene ($\geq \text{C}_3$), C_2H_6 , C_3H_8 , and HCHO. BVOC emissions included isoprene, acetone, acetaldehyde, and lumped alkene ($\geq \text{C}_3$). In the model, the total amount of AVOC emissions was consistent throughout the normal and drought conditions while that of BVOC emissions increased by 65.55 % under the drought condition compared to the normal condition. Therefore, AVOC emission changes could be ruled out and it is the increase of BVOC emissions that caused the large increase of HCHO columns in GEOS-Chem under drought condition. Since the model has a large overestimation of HCHO columns under drought, BVOC emissions were likely overestimated under drought condition. Wildfires are another source that can cause HCHO column bias in the model (Alvarado et al., 2020; Liao et al., 2021). However, most of the wildfires in South Korea occur in the winter-spring season and there was no major wildfire in South Korea in the study period (2016 – 2018 JJA) (Jo et al., 2023).

Given that isoprene emissions were the most dominant BVOC emissions (more than 85 %) in South Korea (Fig. S6), we examined isoprene emissions in the standard GEOS-Chem model by comparing their spatial distributions under the normal and drought conditions (Figs. 3b-d). The reason why isoprene emissions are higher than monoterpenes and sesquiterpenes is because of the type of forests in South Korea. According to the Korea Forest Service (https://english.forest.go.kr/kfswweb/kfi/kfs/cms/cmsView.do?cmsId=FC_001679&mn=UENG_01_03#:~:text=Status%20of%20Forest%20in%20Korea&text=The%20forested%20area%20in%20Korea,times%20higher%20than%20in%201953), about 60% of forests in Korea consist of deciduous-leaved forests (*Quercus mongolica*, *Quercus variabilis*, and *Quercus acutissima*) and mixed forests, and 36.9% are coniferous forests. The deciduous-leaved trees are well-known sources of biogenic isoprene. The mean (total) fluxes of isoprene emissions over South Korea were $3.76 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (25.08 Gg/week) under the normal condition and $6.45 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (42.47 Gg/week) under the drought condition. Notably, the mean (total)

fluxes of isoprene emissions under the drought condition were 71.5 % (68.54 %) higher compared to the normal condition. This increase of isoprene emissions was larger than that of the simulated HCHO column. The difference could be caused by enhanced HCHO loss by the increased photolysis under the clear sky during the drought condition (Naimark et al., 2021; Wang et al., 2022b). Spatially, the increase of isoprene emissions is the largest over the eastern and central parts of the region (Fig. 3d). These areas correspond to two mountain ranges in South Korea, the TaeBaek (TB) and the SoBaek (SB) mountain ranges, respectively (Yu et al., 2023). The high isoprene emissions in these regions are due to the high densities of broadleaf deciduous temperate trees, which are well-known sources of biogenic isoprene (Jang et al., 2020; Yu et al., 2023).

Despite the significant increase of isoprene emissions in the mountain ranges under the drought condition, the increase of the HCHO column under the drought condition was relatively homogeneous in the model (Fig. 2f). The HCHO production depends on the abundance of oxidants such as nitrogen oxides ($\text{NO}_x = \text{NO}_2 + \text{NO}$). In the low NO_x environment, high isoprene emissions can suppress hydroxyl radicals (OH), leading to lower HCHO yield than in the high NO_x environment (Marais et al., 2012; Wells et al., 2020). Indeed, we computed a ratio between the modeled HCHO column change (Drought – Normal) and the modeled isoprene emissions change (Fig. S7a) and found that the ratio tended to be higher over the regions with higher NO_2 column (northwest and southeast) compared to the mountain ranges with lower NO_2 column (Figs. S7b-c). Therefore, the different HCHO productions under the different abundance of NO_x could result in the relatively homogeneous distribution of the modeled HCHO column compared to that of isoprene emissions. Given that the modeled HCHO column was affected by the abundance of NO_x , the model performance on NO_2 could play a role on the larger HCHO column bias under the drought condition. However, GEOS-Chem was found to have only small NO_2 column bias (less than -3.0%) in the South Korean region and the bias was smaller under the drought condition than the normal condition (Fig. S7d). Therefore, larger HCHO column bias under the drought condition was unlikely caused by NO_2 bias.

Based on the above discussions, the larger HCHO column bias under the drought condition was likely caused by the model overestimation (71.5 % increase) of isoprene emissions under the drought condition in South Korea. As described in Section 2.4, MEGAN2.1 implementation in the standard GEOS-Chem model does not have the soil moisture parameter and hence overestimates isoprene emissions under the drought condition. The drought stress algorithms (i.e., WD and JD; Section 2.4) were developed to address this caveat. For example, Wang et al. (2022b) showed that the increases of isoprene emissions under the SE US drought condition were 22.7 – 56 % in GEOS-Chem and those were reduced to 15.2 – 35.23 % after implementing their drought stress algorithm. Given this, we implemented two existing drought stress algorithms (WD and JD; Section 2.4) in the following section and investigated whether they could reduce the overestimation of isoprene emissions and correct the HCHO column bias under the drought condition in South Korea.

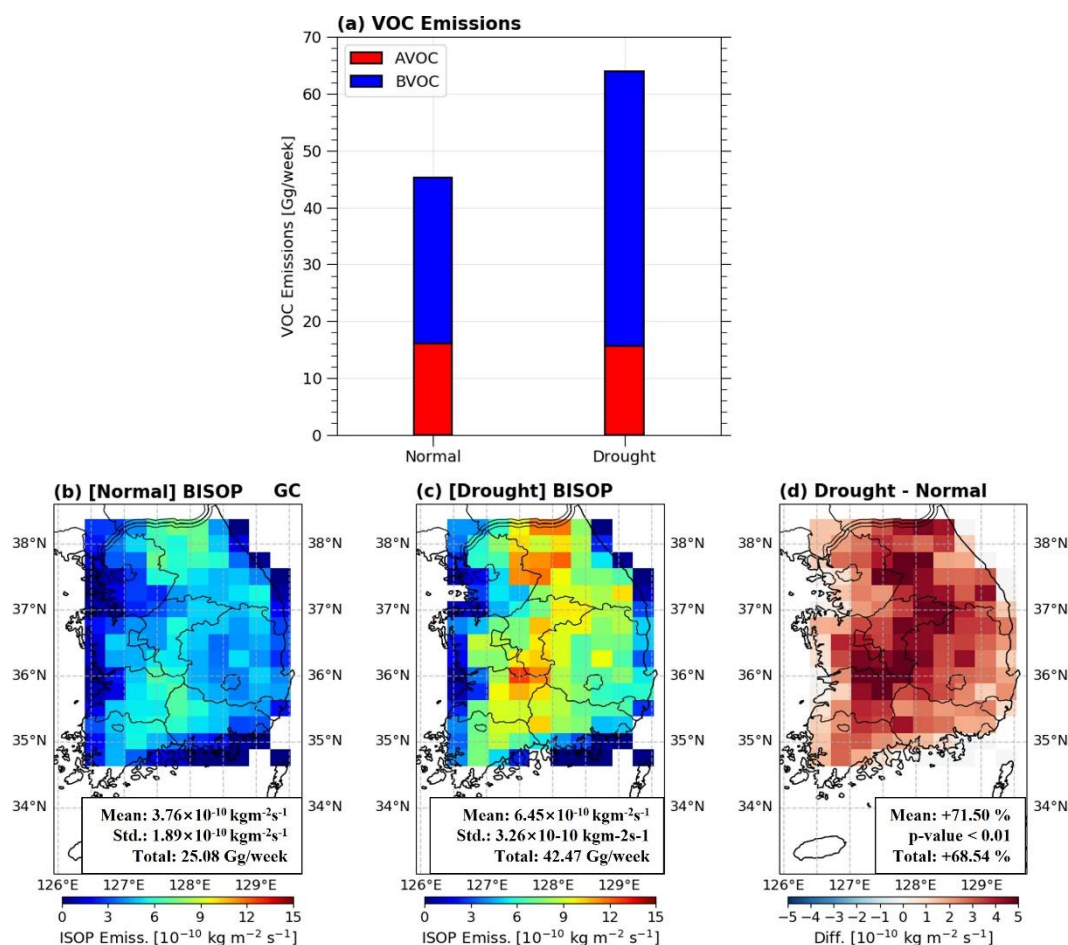


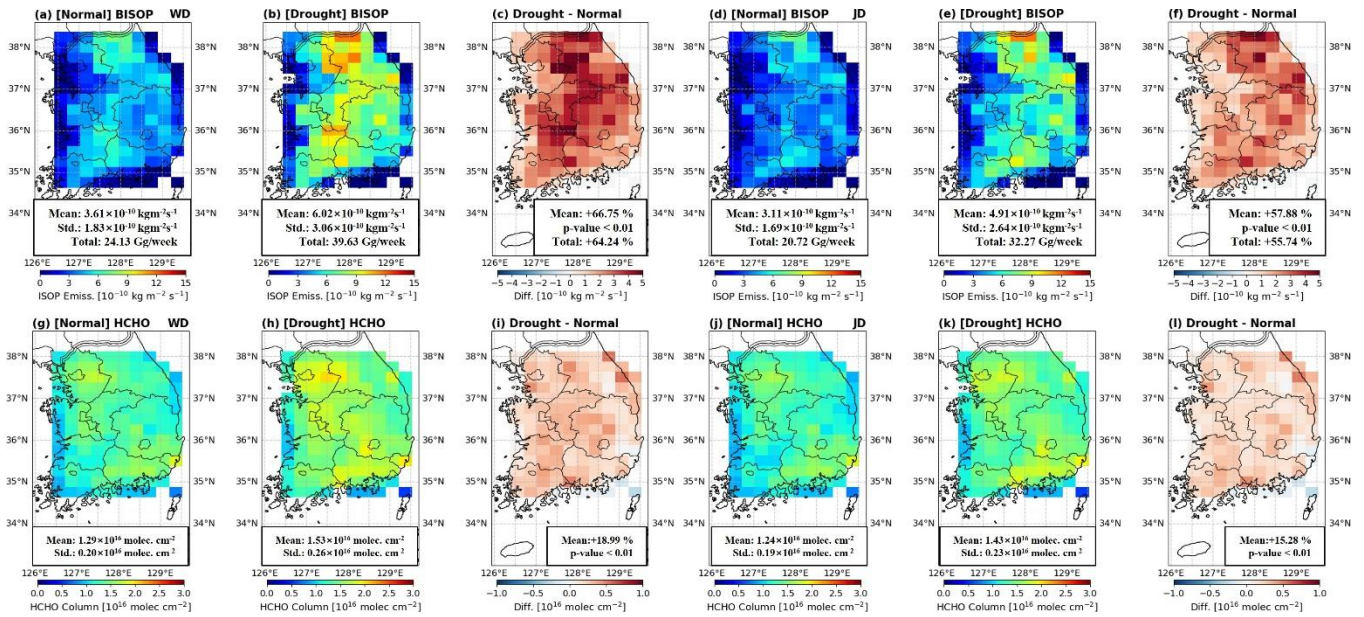
Figure 3: (a) The total amounts of AVOC and BVOC emissions over South Korea under the normal and drought conditions. (b) The biogenic isoprene emissions (BISOP) under (b) the normal condition, (c) the drought condition, and (d) the difference (drought – normal) in the standard GEOS-Chem. The mean and standard deviation (Std.) is presented in each panel and p-value for the mean difference was derived from Student's *t*-test.

3.2 HCHO column biases in two existing drought stress algorithms

Figures 4a-f show the spatial distributions of isoprene emissions simulated by WD and JD. Both WD and JD showed significant increases of isoprene emissions over the mountain ranges under the drought condition, which was consistent with the standard GEOS-Chem, but the magnitudes of the increases were reduced. In WD (Figs. 4a-c), the mean (total) fluxes of isoprene emissions were $3.61 \times 10^{-10} \text{ kg m}^{-2} \text{ s}^{-1}$ (24.13 Gg/week) under the normal condition and $6.02 \times 10^{-10} \text{ kg m}^{-2} \text{ s}^{-1}$ (39.63 Gg/week) under the drought condition. Compared to the standard GEOS-Chem (Table 2), under the normal condition, the mean flux of

isoprene emissions was reduced by $0.15 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (3.9 %) and the total isoprene emission was reduced by 0.95 Gg/week (3.78 %). Under the drought condition (Table 2), the mean flux of isoprene emissions was reduced by $0.43 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (6.67 %) and the total isoprene emission was reduced by 2.84 Gg/week (6.69 %). The increase of the mean (total) fluxes of isoprene emissions under the drought condition was 66.75 % (64.24 %), which was reduced by 4.3 % compared to the standard GEOS-Chem. In JD (Figs. 4d-f), the mean (total) fluxes of isoprene emissions were $3.11 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (20.72 Gg/week) under the normal condition and $4.91 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (32.27 Gg/week) under the drought condition. Compared to the standard GEOS-Chem (Table 2), under the normal condition, the mean flux of isoprene emissions was reduced by $0.65 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (17.29 %) and the total isoprene emission was reduced by 4.36 Gg/week (17.38 %). Under the drought condition (Table 2), the mean flux of isoprene emissions was reduced by $1.54 \times 10^{-10} \text{ kgm}^{-2}\text{s}^{-1}$ (23.88 %) and the total isoprene emission was reduced by 10.20 Gg/week (24.02 %). The increase of the mean (total) fluxes of isoprene emissions under the drought condition was 57.88 % (55.74 %), which was reduced by 12.8 % compared to the standard GEOS-Chem.

The spatial distributions of the HCHO columns in WD and JD are presented in Figures 4g-l. In WD, the mean HCHO columns in South Korea were $1.29 \times 10^{16} \text{ molec. cm}^{-2}$ under the normal condition and $1.53 \times 10^{16} \text{ molec. cm}^{-2}$ under the drought condition (Figs. 4g-i). In JD, they were $1.24 \times 10^{16} \text{ molec. cm}^{-2}$ under the normal condition and $1.43 \times 10^{16} \text{ molec. cm}^{-2}$ under the drought condition (Figs. 4j-l). The mean HCHO column biases (relative bias) under the normal and drought conditions were $0.18 \times 10^{16} \text{ molec. cm}^{-2}$ (16.22 %) and $0.36 \times 10^{16} \text{ molec. cm}^{-2}$ (30.77 %) in WD, respectively, and they were $0.13 \times 10^{16} \text{ molec. cm}^{-2}$ (11.71 %) and $0.26 \times 10^{16} \text{ molec. cm}^{-2}$ (22.22 %) in JD, respectively (Table 1). Although WD and JD reduced the biases of the HCHO column in the standard GEOS-Chem under both normal and drought conditions, the biases were still significant in those simulations especially under the drought condition. Furthermore, both drought stress algorithms simulated a much larger increase of the HCHO columns between normal and drought conditions (18.99 % in WD and 15.28 % in JD) than that indicated by OMI (5.4 %). Therefore, neither WD nor JD was effective for improving isoprene emissions under drought condition in South Korea. This is because both algorithms were developed and tuned based on the SE US region as described in Section 2.4. The two regions have different main deciduous-leaf tree species. South Korea has mainly *Quercus mongolica*, *Quercus variabilis*, and *Quercus acutissima* (Lee et al., 2025), while the SE US has *Quercus stellata*, *Quercus alba*, and *Quercus prinus* (Perry et al., 2022). This fundamental difference may cause ineffectiveness of WD and JD in South Korea. For example, the formula for WD was derived to minimize the HCHO column biases in the SE US where the biases of the HCHO column under drought condition were effectively reduced from $0.08 \times 10^{16} - 0.15 \times 10^{16} \text{ molec. cm}^{-2}$ to $-0.05 \times 10^{16} - 0.02 \times 10^{16} \text{ molec. cm}^{-2}$ by implementing WD (Wang et al., 2022b). In South Korea, however, the bias of the HCHO column under the drought condition was reduced from $0.42 \times 10^{16} \text{ molec. cm}^{-2}$ only to $0.36 \times 10^{16} \text{ molec. cm}^{-2}$ by WD, indicating that the bias of the HCHO column was not effectively reduced by WD.



305 **Figure 4: The biogenic isoprene emissions (BISOP) in WD under (a) the normal condition, (b) the drought condition, and (c) the**
difference (drought – normal). (d-f) Same as a-c but in JD. The HCHO column in WD under (g) the normal condition, (h) the drought
condition, and (i) the difference (drought – normal). (j-l) Same as g-i but in JD. The mean and standard deviation (Std.) is presented
in each panel and p-value for the mean difference was derived from Student’s t-test.

310 **Table 2: The differences (relative difference) in the biogenic isoprene emissions and the HCHO column simulated by each drought**
stress algorithms compared to the standard GEOS-Chem.

Drought stress algorithms	Drought	Mean flux of isoprene emissions	Total isoprene emissions	Mean HCHO column
		[Unit: $10^{-10} \text{ kg m}^{-2} \text{ s}^{-1}$]	[Unit: Gg/week]	[Unit: $10^{16} \text{ molec. cm}^{-2}$]
WD	Normal	-0.15 (-3.9 %)	-0.95 (-3.78 %)	-0.04 (-3.00 %)
	Drought	-0.43 (-6.67 %)	-2.84 (-6.69 %)	-0.06 (-3.77 %)
JD	Normal	-0.65 (-17.29 %)	-4.36 (-17.38 %)	-0.09 (-6.77 %)
	Drought	-1.54 (-23.88 %)	-10.20 (-24.02 %)	-0.16 (-10.06 %)

4. Estimation of isoprene emissions by applying an iterative finite difference mass balance (IFDMB) method

4.1 Iterative finite difference mass balance (IFDMB) method

315 Since the overestimation of isoprene emissions under the drought condition still existed with these two drought stress algorithms, alternative approaches are needed to reduce this overestimation. An inverse modeling method using the satellite HCHO column as top-down constraint could be a useful tool to estimate isoprene emissions. In this section, we applied one of the most widely used inverse modeling methods, an iterative finite difference mass balance (IFDMB) method, to quantify isoprene emissions more accurately and to improve the HCHO column biases in the South Korean region.

320 The IFDMB method is one of the widely used inverse modeling methods to estimate the targeted emissions at the surface from the satellite observation of the associated atmospheric chemical species (Choi et al., 2022; Cooper et al., 2017; Li et al., 2019; Momeni et al., 2024). This study focuses on estimating isoprene emissions by utilizing the HCHO column data derived from OMI satellite observations and GEOS-Chem model simulations. The IFDMB method calculates a scaling factor by assessing the difference between the observed and modeled HCHO columns, incorporating the sensitivity of the modeled HCHO
325 columns to isoprene emissions in each iteration. This scaling factor is subsequently applied to refine isoprene emissions. Based on its simplicity, the IFDMB method is less computationally demanding and can be more efficient (Li et al., 2019) compared to other inverse modeling methods such as a four-dimensional variational data assimilation (4D-Var) method or a hybrid IFDMB-4DVar (Choi et al., 2022; Cooper et al., 2017). A limitation of the IFDMB method is that it neglects the effect of horizontal transport across the model columns, making it most suitable for chemical species with relatively short lifetimes.
330 The lifetime of the isoprene and the HCHO in the atmosphere is known to be short (only a few hours) (Bates and Jacob, 2019), so we used the IFDMB method in this study taking advantage of its simplicity. The detailed estimation of the isoprene emission at each grid and iteration in the IFDMB method is shown below (Cooper et al., 2017; Li et al., 2019; Momeni et al., 2024):

$$E_t = E_a \left(1 + \frac{1}{S} \frac{\Omega_o - \Omega_a}{\Omega_a} \right) \quad (4)$$

where E_t is the newly estimated isoprene emissions at the present iteration (a posterior), E_a is the isoprene emissions at the
335 previous iteration (a priori), Ω_o is the OMI HCHO column, Ω_a is the HCHO column simulated by GEOS-Chem, and S is the sensitivity of the HCHO column with respect to isoprene emission change. S is defined at each grid and iteration as follows:

$$S = \frac{\Delta\Omega/\Omega}{\Delta E/E} \quad (5)$$

where E is the biogenic isoprene emission, ΔE is a change in the biogenic isoprene emission, Ω is the HCHO column, and $\Delta\Omega$ is a change in the HCHO column. To get the initial value of S , an 20 % increase was applied to isoprene emissions simulated
340 by MEGAN2.1 at each grid (Momeni et al., 2024). The S was obtained in every month in all 3 years. To avoid negative values

in the posterior emissions, the constraint ($\frac{1}{5} \frac{\Omega_o - \Omega_a}{\Omega_a} > -1$) was set so that E_t cannot be negative. In addition, to make consistency with other isoprene emission dataset, the maximum value of E_t at each grid was set to long-term (2005 – 2014) daily maximum isoprene emissions from the GlobEmission data (Bauwens et al., 2016) which is another isoprene emission dataset estimated from the OMI HCHO column retrievals. The iteration continued until the normalized mean difference between E_a and E_t was less than 1 % (Momeni et al., 2024). It required 5–7 iterations for each month to get the final posterior isoprene emissions. At every iteration, GEOS-Chem was set to read the newly updated isoprene emissions. Other biogenic emissions such as Limonene, Ethanol, Acetaldehyde, Acetone, Sesquiterpene, Alkenes, Monoterpenes, and MTPA, and all AVOC emissions were set to the same values as in the standard GEOS-Chem simulation to focus on isoprene emission changes. **The assumption behind the application of IFDMB method was that AVOC emissions used in this study were correct, at least with much higher accuracy than isoprene emissions, which is the main caveat in this study. However, AVOC emissions were localized over the northwestern and southeastern parts of South Korea where major metropolitan areas are located (Fig. S8) while isoprene emissions were distributed over most of South Korea (Figs. 3b-c). Also, AVOC emissions were consistent throughout the normal and drought conditions (Fig. S8 and Fig. 3a), which means that changes in HCHO under the drought condition were caused by the changes in isoprene emissions. Therefore, the impact of AVOC emissions may be localized and spatially distinct from that of isoprene emissions.**

4.2 Estimated isoprene emissions by the IFDMB and improvement in the HCHO column biases

Figures 5a-c show the distributions of the final posterior isoprene emissions estimated by the IFDMB. The overall distribution under the normal condition (Fig. 5a) was characterized by the high isoprene emissions over the TB and SB mountain ranges with high densities of biogenic isoprene sources, which was consistent with that from the standard GEOS-Chem (Fig. 3b). The increase of the isoprene emissions under the drought condition was also dominant over these mountain ranges (Figs. 5b-c). **In both normal and drought conditions, therefore, the spatial distribution of isoprene emissions estimated by the IFDMB could represent higher isoprene emissions over the mountain ranges with large density of broadleaf trees.** However, as expected by the positive HCHO column biases in the standard GEOS-Chem (Fig. 2), the IFDMB estimated the isoprene emissions over most of the South Korean region to be lower (Figs. 5a-b) compared to those in the standard GEOS-Chem (Figs. 3b-c) under both normal and drought conditions. The mean (total) fluxes of isoprene emissions over South Korea were $1.87 \times 10^{-10} \text{ kgm}^{-2} \text{ s}^{-1}$ (12.50 Gg/week) under the normal condition and $2.59 \times 10^{-10} \text{ kgm}^{-2} \text{ s}^{-1}$ (17.03 Gg/week) under the drought condition. Compared to the standard GEOS-Chem, isoprene emissions were reduced by 50.16 % under the normal condition, and by 60 % under the drought condition. The increase of isoprene emissions under the drought condition derived by the IFDMB was 36.24 %, much lower than that in the standard GEOS-Chem (68.54 %) and the two drought stress algorithms (64.24 % in WD, 55.74 % in JD). This amount of the increase of isoprene emissions during drought was consistent with those in other regions such as a 30 % increase in northern China (Wang et al., 2021) and 15 – 35 % increases in SE US region (Wang et al., 2022b).

The spatial distributions of the HCHO column derived from the final posterior isoprene emissions are presented in Figs. 5d-f. The mean HCHO columns in South Korea were 1.12×10^{16} molec. cm^{-2} under the normal condition and 1.24×10^{16} molec. cm^{-2} under the drought condition. The increase of the HCHO column under the drought condition was 10.71 %, comparable to that in OMI and much lower than that in the standard GEOS-Chem (20.23%). Figure 5g compares the HCHO columns under the normal and drought conditions in all simulations with OMI. As expected, the smallest HCHO column bias was found in the IFDMB under both normal and drought conditions. The HCHO column biases of the IFDMB were 0.01×10^{16} molec. cm^{-2} under the normal condition and 0.07×10^{16} molec. cm^{-2} under the drought condition. Under the normal condition, the mean HCHO column bias with respect to OMI was 0.9 % in the IFDMB but the biases were 19.82 %, 16.22 %, and 11.71 % in the standard GEOS-Chem, WD, and JD, respectively. Under the drought condition, the mean HCHO column bias with respect to OMI was 5.98 % in the IFDMB compared to 35.9 %, 30.77 %, and 22.22 % in the standard GEOS-Chem, WD, and JD, respectively. Therefore, by adopting the IFDMB-derived biogenic isoprene emissions, the HCHO column biases were reduced under both normal and drought conditions in the South Korean region. This indicates that the IFDMB is effective to reduce the HCHO column biases by adjusting isoprene emissions than the two existing drought stress algorithms.

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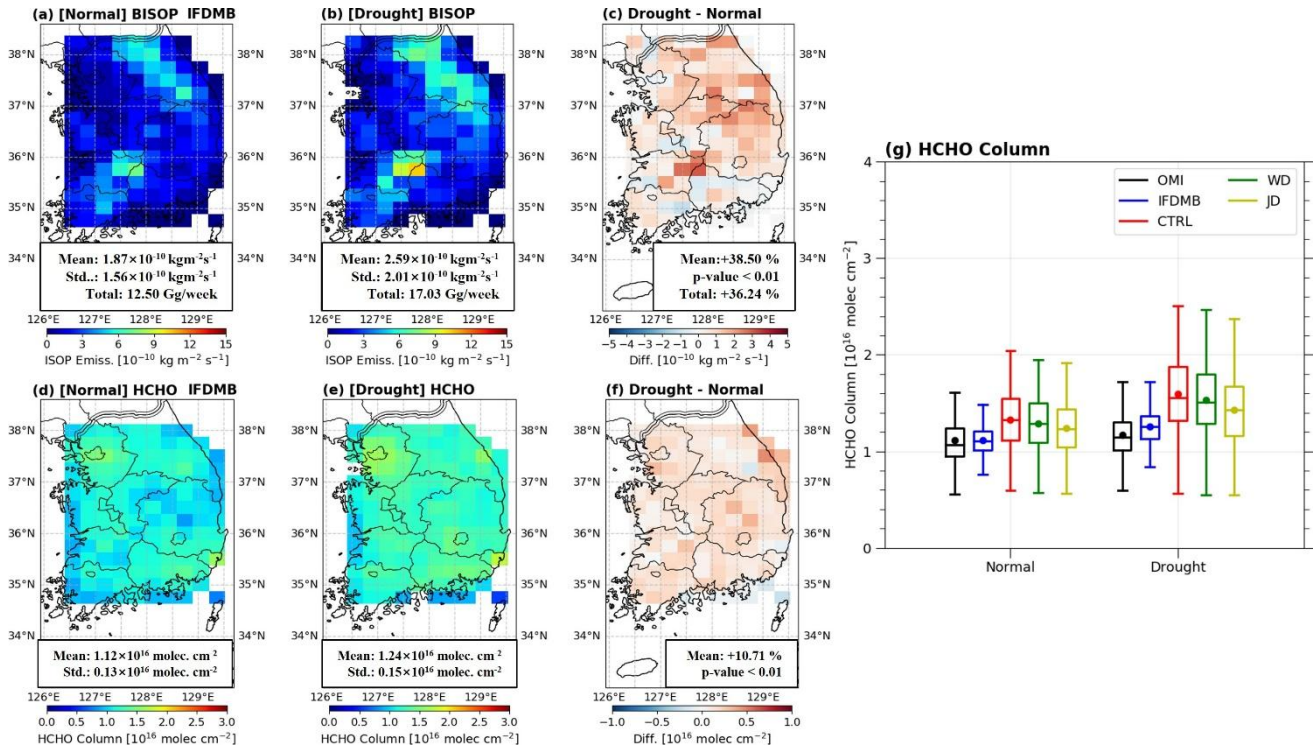


Figure 5: The final posterior biogenic isoprene emissions (BISOP) under (a) the normal condition, (b) the drought condition, and (c) the difference (drought – normal) estimated by the IFDMB. The HCHO column under (d) the normal condition, (e) the drought condition, and (f) the difference (drought – normal) derived from the final posterior BISOP estimated by the IFDMB. **The mean and standard deviation (Std.) is presented in each panel and p-value for the mean difference was derived from Student's *t*-test. (g) The boxplot for HCHO column in OMI and in all GEOS-Chem simulations. The box in boxplot extends from the first quartile (Q1) to the third quartile (Q3) with a line at the median value and a dot at a mean value. The inter-quartile range (IQR) is from Q1 to Q3, and the whisker in boxplot extends from the $Q1-1.5\times IQR$ to the $Q3+1.5\times IQR$.**

4.3 Other secondary aerosols such as O₃ and PM_{2.5} in IFDMB

The changes in isoprene emissions can also lead to changes in other secondary air pollutants such as O₃ and PM_{2.5}. Regarding O₃, previous studies showed that most of the South Korean region is in NO_x saturated (VOC-limited) regime or transitional regime for O₃ formation (Kashfi Yeganeh et al., 2024; Schroeder et al., 2020; Wasti and Wang, 2022). In these regimes, the decrease (increase) in HCHO can lead to decrease (increase) in O₃ concentration. Given that there was no significant difference in NO₂ column between the standard GEOS-Chem and the IFDMB (Fig. S7d), we investigated how the modeled O₃ concentrations changed with posterior isoprene emissions estimated by the IFDMB. Figures 6a-b show scatter plots of daytime (7am – 6pm) mean O₃ concentrations between the surface observations and the model outputs (black for standard GEOS-Chem and red for IFDMB) under the normal condition and the drought condition. Under the normal condition (Fig. 6a), the mean O₃ concentrations in South Korea were 35.50 ppbv and 46.90 ppbv in the observation and the standard GEOS-Chem, respectively. The standard GEOS-Chem had positive O₃ biases in most of the measurement sites, which was indicated by the normalized mean biases (NMB) of 33.58 %. After using posterior isoprene emissions estimated by the IFDMB, the modeled O₃ concentrations decreased in most of the South Korean region. The mean O₃ concentrations in the IFDMB were 44.23 ppbv under the normal condition, indicating that the mean O₃ concentrations were reduced by 2.66 ppbv (5.67 %) with respect to the standard GEOS-Chem by applying IFDMB. As a result, the NMB in IFDMB was 25.91 % under the normal condition, which was reduced by 7.67 % compared to the standard GEOS-Chem. Other metrics such as coefficient of determination (R²) and root mean square error (RMSE) also show improvement in IFDMB compared to the standard GEOS-Chem (Fig. 6a). Under the drought condition (Fig. 6b), the mean observed O₃ concentrations in South Korea was 43.15 ppbv, which was higher than those under the normal condition. The increase in O₃ concentrations under the drought condition was consistent with the expectation of a VOC-limited regime in response to increasing HCHO yet no change in NO₂ under the drought condition as seen by OMI (Fig. S7d). The mean O₃ concentrations in the standard GEOS-Chem was 55.42 ppbv under the drought condition with the NMB of 31.04 %. In IFDMB, the mean O₃ concentrations was 50.47 ppbv under the drought condition, which was reduced by 4.95 ppbv (8.93 %) with respect to the standard GEOS-Chem. This is consistent with the VOC-limited regime where the reduction in isoprene emissions by the IFDMB leads to reduced ozone concentrations. The NMB in the IFDMB was 19.63 % under the drought condition, reduced by 11.41 % compared to the standard GEOS-Chem. As under the normal

condition, the improvement was also indicated by higher R^2 and lower RMSE. Thus, the modeled O_3 concentration was found to be improved by applying the IFDMB method for better isoprene emissions modeling.

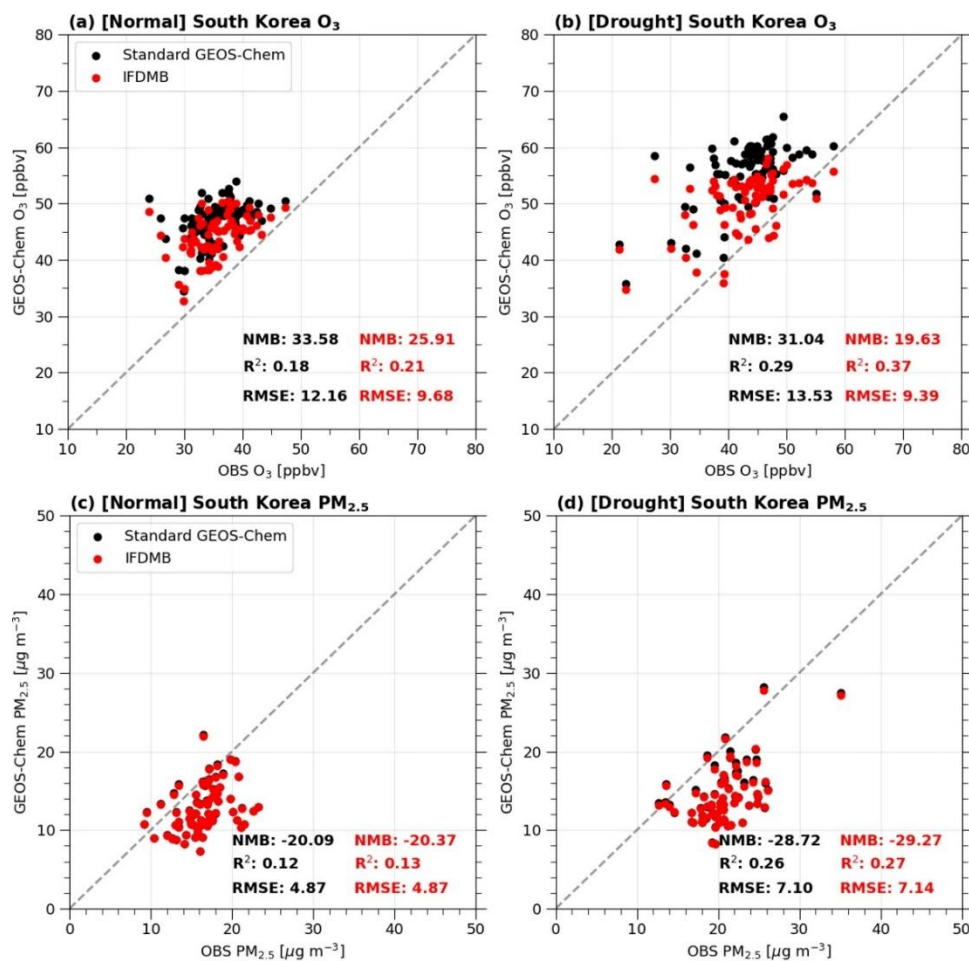


Figure 6: The scatter plot for the observed O_3 and the simulated O_3 under (a) the normal and (b) drought conditions. Black and red dots denote standard GEOS-Chem and IFDMB, respectively. Normalized mean bias (NMB), coefficient of determination (R^2), and root mean square error (RMSE) for the standard GEOS-Chem (black) and IFDMB (red) are presented in each panel. The gray dotted line is 1:1 line. (c-d) Same as a-b but for $PM_{2.5}$.

Figures 6c-d show scatter plots of daytime (7am – 6pm) mean $PM_{2.5}$ concentrations between the surface observations and the model outputs (black for standard GEOS-Chem and red for IFDMB) under the normal condition and the drought condition. Under the normal condition (Fig. 6c), the mean $PM_{2.5}$ concentrations were $16.41 \mu g/m^3$ and $12.42 \mu g/m^3$ in the observation and the standard GEOS-Chem, respectively. The standard GEOS-Chem had negative $PM_{2.5}$ biases in most of the measurement sites except for the Seoul metropolitan area. The NMB for the standard GEOS-Chem was -20.09 %. In IFDMB, contrary to

ozone, the changes in PM_{2.5} concentration were not significant as indicated by the NMB, R², and RMSE. Such insignificant changes in PM_{2.5} in IFDMB were also found under the drought condition (Fig. 6d). In South Korea, the major constituents in PM_{2.5} are the inorganic constituents such as nitrate (NO₃⁻), sulfate (SO₄²⁻), ammonium (NH₄⁺) emitted from the anthropogenic sources (Ryou et al., 2018; Park et al., 2020), which was the reason for the insignificant change in PM_{2.5} between the standard GEOS-Chem and the IFDMB.

4.4 Empirical equations to adjust MEGAN2.1 isoprene emissions in South Korea

Based on the IFDMB results, we derived empirical equations to adjust summertime (JJA) MEGAN2.1 isoprene emissions in South Korea in order to generalize a quantitative estimate of the drought impact on biogenic isoprene emissions in the region. Figure 7a shows the scatterplots between biogenic isoprene emissions scaled by LAI in the standard GEOS-Chem and in IFDMB under both normal and drought conditions. The scaling by the LAI was to consider the regional difference in biogenic isoprene emissions due to the different vegetation coverage (Wang et al., 2022b). We overlaid the surface temperatures at each point because it is known that isoprene emissions generally increase with the surface temperature in MEGAN2.1 (Guenther et al., 2012; Guenther et al., 2006). The significant differences between isoprene emissions in the standard GEOS-Chem and in the IFDMB were found mostly in conditions with high surface temperatures (Figs. 7a-b). This suggests that MEGAN2.1 implemented in the standard GEOS-Chem tends to overestimate isoprene emissions compared to those in the IFDMB in high surface temperatures under both normal and drought conditions. The isoprene emissions in the standard GEOS-Chem have a strong dependence on temperature (Section 2.4). This dependence may be overestimated in South Korea under high temperature conditions.

Given that a lack of ecophysiology module to simulate soil moisture (βt) was the main reason why the standard GEOS-Chem has significant biases in the drought condition, we constructed an equation to adjust MEGAN2.1 isoprene emissions by using soil moisture parameter (βt) to separate the normal and drought conditions in the model and using surface temperature to adjust MEGAN2.1 isoprene emissions in each condition. The basic equation was based on the difference between isoprene emissions from MEGAN2.1 and IFDMB, and was defined as follow:

$$E_{MEGAN} \left(1 + \frac{E_{IFDMB} - E_{MEGAN}}{E_{MEGAN}} \right) = E_{MEGAN} \gamma_{SM_OMI} \quad (6)$$

where E_{MEGAN} is the isoprene emissions in the standard GEOS-Chem, E_{IFDMB} is the isoprene emissions estimated by the IFDMB, and γ_{SM_OMI} is for the term in the parentheses $(1 + \frac{E_{IFDMB} - E_{MEGAN}}{E_{MEGAN}})$ to adjust E_{MEGAN} . The term in the parentheses was referred to as γ_{SM_OMI} because it is derived from OMI and plays the same role as the isoprene emission factors (γ_{SM}) in the WD and JD

460 algorithms (Eq. 2 and Eq. 3) to adjust MEGAN2.1 isoprene emissions. Given that the difference between E_{IFDMB} and E_{MEGAN} was associated with the surface temperatures (Figs. 7a-b), we plotted γ_{SM_OMI} with respect to the surface temperatures under both normal and drought conditions (Figs. 7c-d). It was found that γ_{SM_OMI} exponentially decreased as the surface temperatures increased, which was consistent with the fact that E_{MEGAN} in higher surface temperatures needed a stronger adjustment (Figs. 7a-b). The γ_{SM_OMI} near 295K were close to 1 (Figs. 7c-d), which reflected similar isoprene emissions between the MEGAN2.1 and the IFDMB in lower surface temperature (Figs. 7a-b). Based on the exponentially fitted equations for γ_{SM_OMI} under the normal and drought conditions, the empirical equations for γ_{SM_OMI} were obtained as below:

$$\gamma_{SM_OMI} = \begin{cases} e^{-0.274T} e^{81.00} (\beta_t \geq 0.64, T > 295K): Non - drought \\ e^{-0.166T} e^{49.17} (\beta_t < 0.64, T > 295K): Drought \end{cases} \quad (7)$$

where T is the surface temperature, and the e is the exponential constant. Here, we used the same soil moisture stress threshold (β_t) to separate the normal and drought conditions. Also, we set a temperature threshold to apply these empirical equations only for the surface temperatures above 295K because the fitted equations were obtained based on the samples above 295K. It is noteworthy that the general form of these empirical equations for γ_{SM_OMI} was consistent with the equation in WD, indicating that equations for γ_{SM_OMI} in this study and in WD imposed stronger adjustment to MEGAN2.1 isoprene emissions in higher surface temperatures. However, the magnitude of the adjustment was generally stronger in this study than that in WD given the larger HCHO column biases in WD in South Korea (Figs. 4g-i; Table 1).

475 We further verified the proposed empirical equations for γ_{SM_OMI} (Eq. 7) by applying them to the MEGAN2.1 isoprene emissions and getting the adjusted isoprene emissions (Figs. 7e-f). The overall structures of the adjusted isoprene emissions were comparable to those in IFDMB under both normal and drought conditions (Figs. 5a-b). The pattern correlations between the adjusted isoprene emissions and those in the IFDMB were 0.73 under the normal condition, and 0.80 under the drought condition. The total fluxes of the adjusted isoprene emissions were 13.31 Gg/week the normal condition, 18.52 Gg/week under the drought condition, which was comparable to those in IFDMB (12.5 Gg/week under the normal condition, 17.03 Gg/week under the drought condition). Therefore, it was confirmed that applying the empirical equations for γ_{SM_OMI} to the MEGAN2.1 isoprene emissions could replicate isoprene emissions estimated by the IFDMB without running the inversion.

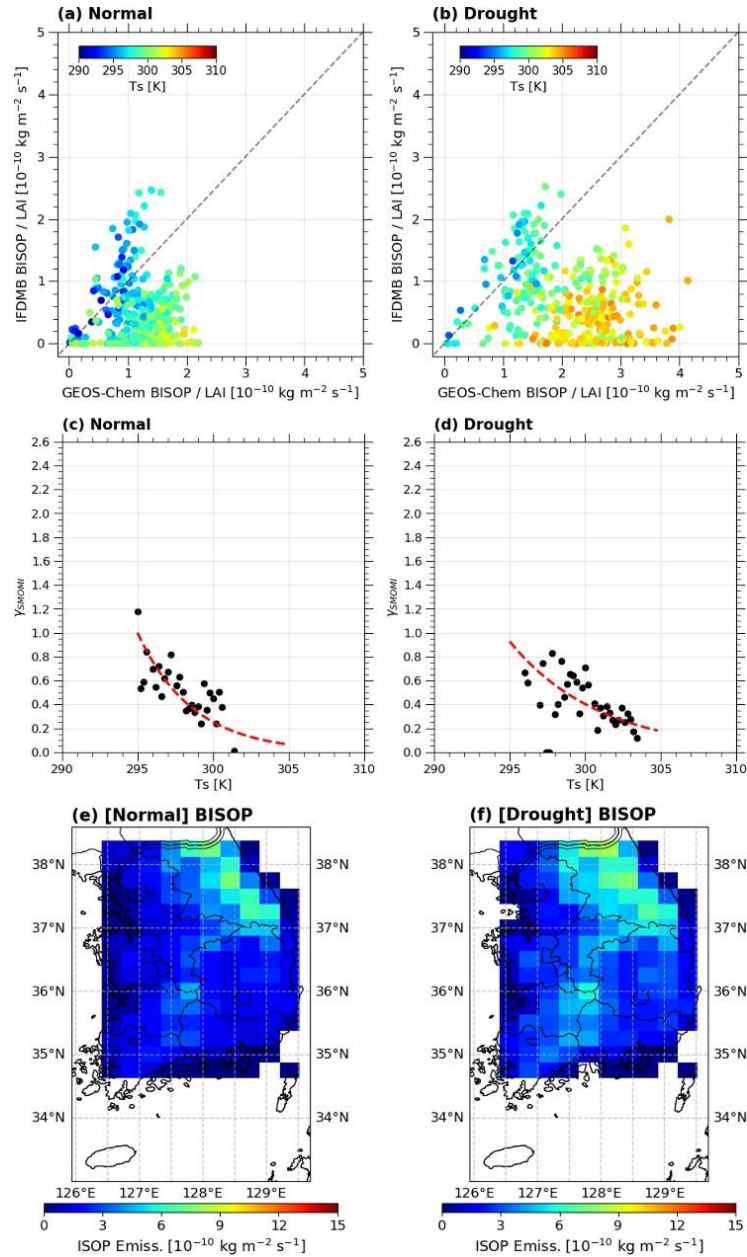


Figure 7: The scatterplots between biogenic isoprene emissions in the standard GEOS-Chem and those in IFDMB under (a) the normal and (b) the drought conditions. Each dot represents isoprene emission value in the standard GEOS-Chem versus that in IFDMB at each grid point in South Korea. Here, the biogenic isoprene emissions were divided by LAI to consider the different vegetation coverage. The surface temperature was overlaid at each grid point. The gray dotted line is 1:1 line. γ_{SM_OMI} with respect to the surface temperatures with an interval of 0.2 K in (c) the normal and (d) the drought conditions. The red dotted lines indicate the fitted lines. The raw data points before binning in 0.2 K are provided in Figure S9. The adjusted biogenic isoprene emissions by the empirical equations for γ_{SM_OMI} under (e) the normal and (f) the drought conditions.

5. Conclusions

In this study, we showed that the mean OMI HCHO columns were 1.11×10^{16} molec. cm^{-2} under the normal condition and 1.17×10^{16} molec. cm^{-2} under the drought condition, indicating 5.4 % increase of the mean HCHO column under the drought condition in South Korea (2016 – 2018 JJA). Compared to the OMI HCHO column, GEOS-Chem overestimated the HCHO column by 19.82 % under the normal condition and by 35.89 % under the drought condition. The worsening performance of GEOS-Chem during drought was likely caused by the overestimation of isoprene emission increase (71.5 %) under the drought condition. Given that MEGAN2.1 implementation in the standard GEOS-Chem model does not have the soil moisture parameter and hence overestimates isoprene emissions under the drought condition, we implemented two existing drought stress algorithms (WD and JD) for the MEGAN2.1 isoprene emission. Although mean flux of isoprene emission was reduced by 3.9 % (17.29 %) under the normal condition and by 6.67 % (23.88%) under the drought condition in WD (JD) compared to the standard GEOS-Chem, GEOS-Chem still overestimated the HCHO column by 16.22 % (11.71%) under the normal condition and by 30.77 % (22.22 %) under the drought condition in WD (JD) compared to the OMI. This result indicated that both WD and JD, which were developed and tuned based on the SE US region, were not effective to reduce HCHO column biases under the drought condition in South Korea.

To improve this, we applied an alternative approach, iterative finite difference mass balance (IFDMB) method, to estimate isoprene emissions based on the HCHO column difference between OMI and the standard GEOS-Chem. By the IFDMB method, the mean fluxes of isoprene emissions were 1.87×10^{-10} $\text{kgm}^{-2} \text{ s}^{-1}$ under the normal condition and 2.59×10^{-10} $\text{kgm}^{-2} \text{ s}^{-1}$ under the drought condition, which were reduced by 50.16 % under the normal condition and by 60 % under the drought condition compared to the standard GEOS-Chem. The mean HCHO columns were 1.12×10^{16} molec. cm^{-2} under the normal condition and 1.24×10^{16} molec. cm^{-2} under the drought condition. The HCHO column biases were significantly reduced to 0.9 % under normal condition and to 5.98 % under the drought condition with respect to the OMI. The increases of the HCHO column under the drought condition was 10.71 %, which was also comparable to the OMI.

The daytime (7am – 6pm) mean observed O_3 concentrations in South Korea were 33.56 ppbv under the normal condition and 42.33 ppbv under the drought condition. The standard GEOS-Chem could simulate the increase of O_3 under the drought condition, but it overestimated O_3 by 32.24 % under the normal condition and by 25.36 % under the drought condition. By applying IFDMB, these biases were improved to 24.64 % under the normal condition and to 12.98 % under the drought condition. The $\text{PM}_{2.5}$ changes between the standard GEOS-Chem and IFDMB were insignificant because major constituents in $\text{PM}_{2.5}$ in South Korea was inorganic components emitted from the anthropogenic sources.

We also proposed the empirical equations that could adjust the summertime MEGAN2.1 isoprene emissions in the South Korean region. The empirical equations were constructed based on isoprene emission difference between the standard GEOS-

Chem and the IFDMB. The empirical equations were dependent on the surface temperature because significant isoprene emissions differences were found in higher surface temperatures. The adjusted isoprene emissions with these empirical equations had consistent amounts and spatial structure with those estimated by the IFDMB, which confirmed that these empirical equations could be used in the MEGAN2.1 to adjust the isoprene emissions in South Korea.

The GEOS-Chem in this study was driven by GEOS-FP meteorology dataset that is one of the conventional and widely used meteorology datasets to drive GEOS-Chem. We found that the surface temperatures from GEOS-FP meteorology have positive biases compared to the observed surface temperatures in South Korea (Fig. S10a). Given that MEGAN2.1 isoprene emissions tended to have larger biases in the higher temperature (Figs. 7a-b), these biases in surface temperature might cause the overestimation of isoprene emission in the standard GEOS-Chem under both normal and drought conditions. Indeed, it was found that the regions with larger surface temperature biases tend to have larger reductions of biogenic isoprene emissions by IFDMB (Figs. S10b-c). So, isoprene emissions might be better simulated in MEGAN2.1 with the bias-corrected surface temperature in GEOS-FP meteorology dataset, which needs future investigation. In spite of this, the proposed empirical equations would be useful for the future study investigating drought stress in South Korea using GEOS-FP meteorology datasets.

As presented in this study, the air quality (O_3) worsens under the drought condition in South Korea. Some climate studies have showed that the drought would occur frequently with the future climate change (Zscheischler et al., 2020). So, the exquisite modeling of isoprene emissions under the drought stress is important in the air quality study. To achieve this, in-situ isoprene emission flux measurements in the South Korea, which cover wide regions and long time periods, would be helpful to constrain the future air quality projection.

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Code and data availability

The GEOS-Chem version 12-7-2 is available at <https://zenodo.org/records/3701669>. The information for the ecophysiology module implemented in GEOS-Chem is available at <https://github.com/geoschem/geos-chem/pull/629/files#diff-7df9bad6db453c401ce2a47eb0236a70f9b0951f2fcff1303e39a2cc835580c7>. The DEDI is available at <https://zenodo.org/records/7768534>, and the observed SPI1 in South Korea is available at <https://data.kma.go.kr/resources/html/en/aowdp.html>. The in-situ observation of ozone and $PM_{2.5}$ in South Korea can be obtained from <https://www.airkorea.or.kr/eng/>. The OMI satellite HCHO and NO_2 columns can be obtained from

550 https://disc.gsfc.nasa.gov/datasets/OMHCHOd_003/summary?keywords=omhchod and
https://disc.gsfc.nasa.gov/datasets/OMNO2d_003/summary, respectively.

Author contributions

YW supervised the entire research. YCJ conducted overall analysis and all model simulations using GEOS-Chem version 12-7-2 provided by HK and RJP. WL contributed to implementing the ecophysiology module and two drought stress algorithms
555 to GEOS-Chem. MM contributed to applying the IFDMB method. YCJ wrote the manuscript draft, and all authors contributed to the preparation of the final version of the manuscript.

Competing interests

The authors declare that they have no conflict of interest.

References

- 560 Alvarado, L. M., Richter, A., Vrekoussis, M., Hilboll, A., Kalisz Hedegaard, A. B., Schneising, O., and Burrows, J. P.: Unexpected long-range transport of glyoxal and formaldehyde observed from the Copernicus Sentinel-5 Precursor satellite during the 2018 Canadian wildfires, *Atmos. Chem. Phys.*, 20, 2057-2072, <https://doi.org/10.5194/acp-20-2057-2020>, 2020.
- Atkinson, R.: Atmospheric chemistry of VOCs and NO_x, *Atmos. Environ.*, 34, 2063-2101, [https://doi.org/10.1016/S1352-2310\(99\)00460-4](https://doi.org/10.1016/S1352-2310(99)00460-4), 2000.
- 565 Bates, K. H. and Jacob, D. J.: A new model mechanism for atmospheric oxidation of isoprene: global effects on oxidants, nitrogen oxides, organic products, and secondary organic aerosol, *Atmos. Chem. Phys.*, 19, 9613-9640, <https://doi.org/10.5194/acp-19-9613-2019>, 2019.
- Bauwens, M., Stavrakou, T., Müller, J.-F., De Smedt, I., Van Roozendaal, M., Van Der Werf, G. R., Wiedinmyer, C., Kaiser, J. W., Sindelarova, K., and Guenther, A.: Nine years of global hydrocarbon emissions based on source inversion of OMI
570 formaldehyde observations, *Atmos. Chem. Phys.*, 16, 10133-10158, <https://doi.org/10.5194/acp-16-10133-2016>, 2016.
- Best, M. J., Pryor, M., Clark, D., Rooney, G. G., Essery, R., Ménard, C., Edwards, J., Hendry, M., Porson, A., and Gedney, N.: The Joint UK Land Environment Simulator (JULES), model description–Part 1: energy and water fluxes, *Geosci. Model Dev.*, 4, 677-699, <https://doi.org/10.5194/gmd-4-677-2011>, 2011.
- Bey, I., Jacob, D. J., Yantosca, R. M., Logan, J. A., Field, B. D., Fiore, A. M., Li, Q., Liu, H. Y., Mickley, L. J., and Schultz,
575 M. G.: Global modeling of tropospheric chemistry with assimilated meteorology: Model description and evaluation, *J. Geophys. Res.*, 106, 23073-23095, <https://doi.org/10.1029/2001JD000807>, 2001.
- Chance, K.: OMI/Aura Formaldehyde (HCHO) Total Column Daily L3 Weighted Mean Global 0.1 deg Lat/Lon Grid V003,

- Greenbelt, MD, USA, Goddard Earth Sciences Data and Information Services Center (GES DISC) [dataset], 2019.
- Chen, D., Wang, Y., McElroy, M. B., He, K., Yantosca, R. M., and Le Sager, P.: Regional CO pollution and export in China
 580 simulated by the high-resolution nested-grid GEOS-Chem model, *Atmos. Chem. Phys.*, 9, 3825-3839,
<https://doi.org/10.5194/acp-9-3825-2009>, 2009.
- Choi, J., Henze, D. K., Cao, H., Nowlan, C. R., González Abad, G., Kwon, H. A., Lee, H. M., Oak, Y. J., Park, R. J., and Bates,
 K. H.: An inversion framework for optimizing non-methane VOC emissions using remote sensing and airborne
 observations in northeast Asia during the KORUS-AQ field campaign, *J. Geophys. Res.*, 127, e2021JD035844,
 585 <https://doi.org/10.1029/2021JD035844>, 2022.
- Clark, D., Mercado, L., Sitch, S., Jones, C., Gedney, N., Best, M., Pryor, M., Rooney, G., Essery, R., and Blyth, E.: The Joint
 UK Land Environment Simulator (JULES), model description–Part 2: carbon fluxes and vegetation dynamics, *Geosci.*
Model Dev., 4, 701-722, <https://doi.org/10.5194/gmd-4-701-2011>, 2011.
- Cooper, M., Martin, R. V., Padmanabhan, A., and Henze, D. K.: Comparing mass balance and adjoint methods for inverse
 590 modeling of nitrogen dioxide columns for global nitrogen oxide emissions, *J. Geophys. Res.*, 122, 4718-4734,
<https://doi.org/10.1002/2016JD025985>, 2017.
- Ferracci, V., Bolas, C. G., Freshwater, R. A., Staniaszek, Z., King, T., Jaars, K., Otu-Larbi, F., Beale, J., Malhi, Y., and Waite,
 T. W.: Continuous isoprene measurements in a UK temperate forest for a whole growing season: Effects of drought stress
 during the 2018 heatwave, *Geophys. Res. Lett.*, 47, e2020GL088885, <https://doi.org/10.1029/2020GL088885>, 2020.
- 595 Fisher, J. A., Jacob, D. J., Travis, K. R., Kim, P. S., Marais, E. A., Chan Miller, C., Yu, K., Zhu, L., Yantosca, R. M., and
 Sulprizio, M. P.: Organic nitrate chemistry and its implications for nitrogen budgets in an isoprene-and monoterpene-rich
 atmosphere: constraints from aircraft (SEAC 4 RS) and ground-based (SOAS) observations in the Southeast US, *Atmos.*
Chem. Phys., 16, 5969-5991, <https://doi.org/10.5194/acp-16-5969-2016>, 2016.
- Fountoukis, C. and Nenes, A.: ISORROPIA II: a computationally efficient thermodynamic equilibrium model for $K^+ - Ca^{2+} -$
 600 $Mg^{2+} - NH_4^+ - Na^+ - SO_4^{2-} - NO_3^- - Cl^- - H_2O$ aerosols, *Atmos. Chem. Phys.*, 7, 4639-4659,
<https://doi.org/10.5194/acp-7-4639-2007>, 2007.
- Funk, J. L., Jones, C. G., Gray, D. W., Throop, H. L., Hyatt, L. A., and Lerdau, M. T.: Variation in isoprene emission from
Quercus rubra: Sources, causes, and consequences for estimating fluxes, *J. Geophys. Res.*, 110,
<https://doi.org/10.1029/2004JD005229>, 2005.
- 605 Guenther, A., Karl, T., Harley, P., Wiedinmyer, C., Palmer, P. I., and Geron, C.: Estimates of global terrestrial isoprene
 emissions using MEGAN (Model of Emissions of Gases and Aerosols from Nature), *Atmos. Chem. Phys.*, 6, 3181-3210,
<https://doi.org/10.5194/acp-6-3181-2006>, 2006.
- Guenther, A., Jiang, X., Heald, C. L., Sakulyanontvittaya, T., Duhl, T. a., Emmons, L., and Wang, X.: The Model of Emissions
 of Gases and Aerosols from Nature version 2.1 (MEGAN2. 1): an extended and updated framework for modeling biogenic
 610 emissions, *Geosci. Model Dev.*, 5, 1471-1492, <https://doi.org/10.5194/gmd-5-1471-2012>, 2012.

- Hersbach, H., Bell, B., Berrisford, P., Hirahara, S., Horányi, A., Muñoz-Sabater, J., Nicolas, J., Peubey, C., Radu, R., and Schepers, D.: The ERA5 global reanalysis, *Q. J. R. Meteorol. Soc.*, 146, 1999-2049, <https://doi.org/10.1002/qj.3803>, 2020.
- Holm, S. M. and Balmes, J. R.: Systematic review of ozone effects on human lung function, 2013 through 2020, *Chest*, 161, 190-201, <https://doi.org/10.1016/j.chest.2021.07.2170>, 2022.
- 615 Huang, L., McGaughey, G., McDonald-Buller, E., Kimura, Y., and Allen, D. T.: Quantifying regional, seasonal and interannual contributions of environmental factors on isoprene and monoterpene emissions estimates over eastern Texas, *Atmos. Environ.*, 106, 120-128, <https://doi.org/10.1016/j.atmosenv.2015.01.072>, 2015.
- Jang, Y., Eo, Y., Jang, M., Woo, J.-H., Kim, Y., Lee, J.-B., and Lim, J.-H.: Impact of land cover and leaf area index on BVOC emissions over the Korean Peninsula, *Atmosphere*, 11, 806, <https://doi.org/10.3390/atmos11080806>, 2020.
- 620 Jiang, X., Guenther, A., Potosnak, M., Geron, C., Seco, R., Karl, T., Kim, S., Gu, L., and Pallardy, S.: Isoprene emission response to drought and the impact on global atmospheric chemistry, *Atmos. Environ.*, 183, 69-83, <https://doi.org/10.1016/j.atmosenv.2018.01.026>, 2018.
- Jo, H.-W., Krasovskiy, A., Hong, M., Corning, S., Kim, W., Kraxner, F., and Lee, W.-K.: Modeling Historical and Future Forest Fires in South Korea: The FLAM Optimization Approach, *Remote Sens.*, 15, 1446, <https://doi.org/10.3390/rs15051446>, 2023.
- 625 Kashfi Yeganeh, A., Momeni, M., Choi, Y., Park, J., and Jung, J.: A case study of surface ozone source contributions in the Seoul metropolitan area using the adjoint of CMAQ, *J. Air Waste Manag. Assoc.*, <https://doi.org/10.1080/10962247.2024.2361021>, 2024.
- Kim, H.-K., Song, C.-K., Han, K. M., Eo, Y. D., Song, C. H., Park, R., Hong, S.-C., Kim, S.-K., and Woo, J.-H.: Impact of biogenic emissions on early summer ozone and fine particulate matter exposure in the Seoul Metropolitan Area of Korea, *Air Qual. Atmos. Health.*, 11, 1021-1035, <https://doi.org/10.1007/s11869-018-0602-4>, 2018.
- 630 Kim, S.-Y., Jiang, X., Lee, M., Turnipseed, A., Guenther, A., Kim, J.-C., Lee, S.-J., and Kim, S.: Impact of biogenic volatile organic compounds on ozone production at the Taehwa Research Forest near Seoul, South Korea, *Atmos. Environ.*, 70, 447-453, <https://doi.org/10.1016/j.atmosenv.2012.11.005>, 2013.
- 635 Kwon, H.-A., Park, R. J., Oak, Y. J., Nowlan, C. R., Janz, S. J., Kowalewski, M. G., Fried, A., Walega, J., Bates, K. H., Choi, J., Blake, D. R., Wisthaler, A., and Woo, J.-H.: Top-down estimates of anthropogenic VOC emissions in South Korea using formaldehyde vertical column densities from aircraft during the KORUS-AQ campaign, *Elem Sci Anth*, 9, 00109, <https://doi.org/10.1525/elementa.2021.00109>, 2021.
- Lam, J. C. Y., Tai, A. P. K., Ducker, J. A., and Holmes, C. D.: Development of an ecophysiology module in the GEOS-Chem chemical transport model version 12.2.0 to represent biosphere-atmosphere fluxes relevant for ozone air quality, *Geosci. Model Dev.*, 16, 2323-2342, <https://doi.org/10.5194/gmd-16-2323-2023>, 2023.
- 640 Lee, H.-M. and Park, R. J.: Factors determining the seasonal variation of ozone air quality in South Korea: Regional background versus domestic emission contributions, *Environ. Pollut.*, 308, 119645,

<https://doi.org/10.1016/j.envpol.2022.119645>, 2022.

- 645 Lee, J., Yang, T., and Choi, C.: Mapping Nine Dominant Tree Species in the Korean Peninsula Using U-Net and Harmonic Analysis of Sentinel-2 Imagery. *Korean J. Remote Sens.* 41, 243-260, <https://doi.org/10.7780/kjrs.2025.41.2.1.1>, 2025.
- Lee, K.-Y., Kwak, K.-H., Ryu, Y.-H., Lee, S.-H., and Baik, J.-J.: Impacts of biogenic isoprene emission on ozone air quality in the Seoul metropolitan area, *Atmos. Environ.*, 96, 209-219, <https://doi.org/10.1016/j.atmosenv.2014.07.036>, 2014.
- Li, C., Martin, R. V., Shephard, M. W., Cady-Pereira, K., Cooper, M. J., Kaiser, J., Lee, C. J., Zhang, L., and Henze, D. K.:
650 Assessing the iterative finite difference mass balance and 4D-Var methods to derive ammonia emissions over North America using synthetic observations, *J. Geophys. Res.*, 124, 4222-4236, <https://doi.org/10.1029/2018JD030183>, 2019.
- Li, W., Wang, Y., Flynn, J., Griffin, R. J., Guo, F., and Schnell, J. L.: Spatial variation of surface O₃ responses to drought over the contiguous United States during summertime: Role of precursor emissions and ozone chemistry, *J. Geophys. Res.*, 127, e2021JD035607, <https://doi.org/10.1029/2021JD035607>, 2022.
- 655 Liao, J., Wolfe, G. M., Hannun, R. A., St. Clair, J. M., Hanisco, T. F., Gilman, J. B., Lamplugh, A., Selimovic, V., Diskin, G. S., and Nowak, J. B.: Formaldehyde evolution in US wildfire plumes during the Fire Influence on Regional to Global Environments and Air Quality experiment (FIREX-AQ), *Atmos. Chem. Phys.*, 21, 18319-18331, <https://doi.org/10.5194/acp-21-18319-2021>, 2021.
- Limousin, J.-M., Misson, L., Lavoit, A.-V., Martin, N. K., and Rambal, S.: Do photosynthetic limitations of evergreen *Quercus*
660 *ilex* leaves change with long-term increased drought severity?, *Plant Cell Environ.*, 33, 863-875, <https://doi.org/10.1111/j.1365-3040.2009.02112.x>, 2010.
- Mao, J., Fan, S., Jacob, D. J., and Travis, K. R.: Radical loss in the atmosphere from Cu-Fe redox coupling in aerosols, *Atmos. Chem. Phys.*, 13, 509-519, <https://doi.org/10.5194/acp-13-509-2013>, 2013.
- Marais, E. A., Jacob, D. J., Jimenez, J. L., Campuzano-Jost, P., Day, D. A., Hu, W., Krechmer, J., Zhu, L., Kim, P. S., and
665 Miller, C. C.: Aqueous-phase mechanism for secondary organic aerosol formation from isoprene: application to the southeast United States and co-benefit of SO₂ emission controls, *Atmos. Chem. Phys.*, 16, 1603-1618, <https://doi.org/10.5194/acp-16-1603-2016>, 2016.
- Marais, E. A., Jacob, D. J., Kurosu, T., Chance, K., Murphy, J., Reeves, C., Mills, G., Casadio, S., Millet, D., and Barkley, M. P.: Isoprene emissions in Africa inferred from OMI observations of formaldehyde columns, *Atmos. Chem. Phys.*, 12, 6219-
670 6235, <https://doi.org/10.5194/acp-12-6219-2012>, 2012.
- McKee, T. B., Doesken, N. J., and Kleist, J.: The relationship of drought frequency and duration to time scales, *Proceedings of the 8th Conference on Applied Climatology*, 179-183,
- Momeni, M., Choi, Y., Yeganeh, A. K., Pouyaei, A., Jung, J., Park, J., Shephard, M. W., Dammers, E., and Cady-Pereira, K. E.: Constraining East Asia ammonia emissions through satellite observations and iterative finite difference mass balance
675 (iFDMB) and investigating its impact on inorganic fine particulate matter, *Environ. Int.*, 108473, <https://doi.org/10.1016/j.envint.2024.108473>, 2024.

- Morfopoulos, C., Müller, J. F., Stavrakou, T., Bauwens, M., De Smedt, I., Friedlingstein, P., Prentice, I. C., and Regnier, P.: Vegetation responses to climate extremes recorded by remotely sensed atmospheric formaldehyde, *Glob. Change Biol.*, 28, 1809-1822, <https://doi.org/10.1111/gcb.15880>, 2022.
- 680 Müller, J. F., Stavrakou, T., Oomen, G. M., Opacka, B., De Smedt, I., Guenther, A., Vigouroux, C., Langerock, B., Aquino, C. A. B., Grutter, M., Hannigan, J., Hase, F., Kivi, R., Lutsch, E., Mahieu, E., Makarova, M., Metzger, J. M., Morino, I., Murata, I., Nagahama, T., Notholt, J., Ortega, I., Palm, M., Röhling, A., Stremme, W., Strong, K., Sussmann, R., Té, Y., and Fried, A.: Bias correction of OMI HCHO columns based on FTIR and aircraft measurements and impact on top-down emission estimates, *Atmos. Chem. Phys.*, 24, 2207-2237, <https://doi.org/10.5194/acp-24-2207-2024>, 2024.
- 685 Naimark, J. G., Fiore, A. M., Jin, X., Wang, Y., Klovenski, E., and Braneon, C.: Evaluating drought responses of surface ozone precursor proxies: Variations with land cover type, precipitation, and temperature, *Geophys. Res. Lett.*, 48, e2020GL091520, <https://doi.org/10.1029/2020GL091520>, 2021.
- Pacifico, F., Harrison, S., Jones, C., and Sitch, S.: Isoprene emissions and climate, *Atmos. Environ.*, 43, 6121-6135, <https://doi.org/10.1016/j.atmosenv.2009.09.002>, 2009.
- 690 Palmer, P. I., Jacob, D. J., Fiore, A. M., Martin, R. V., Chance, K., and Kurosu, T. P.: Mapping isoprene emissions over North America using formaldehyde column observations from space, *J. Geophys. Res.*, 108, <https://doi.org/10.1029/2002JD002153>, 2003.
- Park, E. H., Heo, J., Kim, H., and Yi, S.-M.: Long term trends of chemical constituents and source contributions of PM_{2.5} in Seoul, *Chemosphere*, 251, 126371, <https://doi.org/10.1016/j.chemosphere.2020.126371>, 2020.
- 695 Pegoraro, E., Rey, A., Greenberg, J., Harley, P., Grace, J., Malhi, Y., and Guenther, A.: Effect of drought on isoprene emission rates from leaves of *Quercus virginiana* Mill, *Atmos. Environ.*, 38, 6149-6156, <https://doi.org/10.1016/j.atmosenv.2004.07.028>, 2004.
- Perry, C. H., Finco, M. V., and Barry, T.: Forest atlas of the United States (U.S. Department of Agriculture, Forest Service, Northern Research Station), Available at: [https://www.fs.usda.gov/sites/default/files/fs_media/fs_document/Forest-Atlas-](https://www.fs.usda.gov/sites/default/files/fs_media/fs_document/Forest-Atlas-of-the-United-States.pdf)
- 700 [of-the-United-States.pdf](https://www.fs.usda.gov/sites/default/files/fs_media/fs_document/Forest-Atlas-of-the-United-States.pdf), 2022.
- Potosnak, M. J., LeSturgeon, L., Pallardy, S. G., Hosman, K. P., Gu, L., Karl, T., Geron, C., and Guenther, A. B.: Observed and modeled ecosystem isoprene fluxes from an oak-dominated temperate forest and the influence of drought stress, *Atmos. Environ.*, 84, 314-322, <https://doi.org/10.1016/j.atmosenv.2013.11.055>, 2014.
- Rissanen, K., Aalto, J., Gessler, A., Hölttä, T., Rigling, A., Schaub, M., and Bäck, J.: Drought effects on volatile organic compound emissions from Scots pine stems, *Plant Cell Environ.*, 45, 23-40, <https://doi.org/10.1111/pce.14219>, 2022.
- 705 Ryou, H., Heo, J., and Kim, S.-Y.: Source apportionment of PM₁₀ and PM_{2.5} air pollution, and possible impacts of study characteristics in South Korea, *Environ. Pollut.*, 240, 963-972, <https://doi.org/10.1016/j.envpol.2018.03.066>, 2018.
- Schroeder, J. R., Crawford, J. H., Ahn, J.-Y., Chang, L., Fried, A., Walega, J., Weinheimer, A., Montzka, D. D., Hall, S. R., and Ullmann, K.: Observation-based modeling of ozone chemistry in the Seoul metropolitan area during the Korea-United

- 710 States Air Quality Study (KORUS-AQ), *Elem. Sci. Anth.*, 8, 3, <https://doi.org/10.1525/elementa.400>, 2020.
- Seco, R., Holst, T., Davie-Martin, C. L., Simin, T., Guenther, A., Pirk, N., Rinne, J., and Rinnan, R.: Strong isoprene emission response to temperature in tundra vegetation, *Proc. Natl. Acad. Sci.*, 119, e2118014119, <https://doi.org/10.1073/pnas.2118014119>, 2022.
- 715 Seco, R., Karl, T., Guenther, A., Hosman, K. P., Pallardy, S. G., Gu, L., Geron, C., Harley, P., and Kim, S.: Ecosystem-scale volatile organic compound fluxes during an extreme drought in a broadleaf temperate forest of the Missouri Ozarks (central USA), *Glob. Change Biol.*, 21, 3657-3674, <https://doi.org/10.1111/gcb.12980>, 2015.
- Shen, L., Jacob, D. J., Zhu, L., Zhang, Q., Zheng, B., Sulprizio, M. P., Li, K., De Smedt, I., González Abad, G., and Cao, H.: The 2005–2016 trends of formaldehyde columns over China observed by satellites: Increasing anthropogenic emissions of volatile organic compounds and decreasing agricultural fire emissions, *Geophys. Res. Lett.*, 46, 4468-4475, <https://doi.org/10.1029/2019GL082172>, 2019.
- 720 Sindelarova, K., Granier, C., Bouarar, I., Guenther, A., Tilmes, S., Stavrakou, T., Müller, J.-F., Kuhn, U., Stefani, P., and Knorr, W.: Global data set of biogenic VOC emissions calculated by the MEGAN model over the last 30 years, *Atmos. Chem. Phys.*, 14, 9317-9341, <https://doi.org/10.5194/acp-14-9317-2014>, 2014.
- Sprengnether, M., Demerjian, K. L., Donahue, N. M., and Anderson, J. G.: Product analysis of the OH oxidation of isoprene and 1, 3-butadiene in the presence of NO, *J. Geophys. Res.*, 107, ACH 8-1-ACH 8-13, <https://doi.org/10.1029/2001JD000716>, 2002.
- 725 Van der Werf, G. R., Randerson, J. T., Giglio, L., Collatz, G., Mu, M., Kasibhatla, P. S., Morton, D. C., DeFries, R., Jin, Y. v., and van Leeuwen, T. T.: Global fire emissions and the contribution of deforestation, savanna, forest, agricultural, and peat fires (1997–2009), *Atmos. Chem. Phys.*, 10, 11707-11735, <https://doi.org/10.5194/acp-10-11707-2010>, 2010.
- 730 Wang, H., Lu, X., Seco, R., Stavrakou, T., Karl, T., Jiang, X., Gu, L., and Guenther, A. B.: Modeling isoprene emission response to drought and heatwaves within MEGAN using evapotranspiration data and by coupling with the community land model, *J. Adv. Model. Earth Syst.*, 14, e2022MS003174, <https://doi.org/10.1029/2022MS003174>, 2022a.
- Wang, P., Liu, Y., Dai, J., Fu, X., Wang, X., Guenther, A., and Wang, T.: Isoprene emissions response to drought and the impacts on ozone and SOA in China, *J. Geophys. Res.*, 126, e2020JD033263, <https://doi.org/10.1029/2020JD033263>, 2021.
- 735 Wang, Y., Lin, N., Li, W., Guenther, A., Lam, J. C., Tai, A. P., Potosnak, M. J., and Seco, R.: Satellite-derived constraints on the effect of drought stress on biogenic isoprene emissions in the southeastern US, *Atmos. Chem. Phys.*, 22, 14189-14208, <https://doi.org/10.5194/acp-22-14189-2022>, 2022b.
- Wang, Y. X., McElroy, M. B., Jacob, D. J., and Yantosca, R. M.: A nested grid formulation for chemical transport over Asia: Applications to CO, *J. Geophys. Res.*, 109, <https://doi.org/10.1029/2004JD005237>, 2004.
- 740 Wasti, S. and Wang, Y.: Spatial and temporal analysis of HCHO response to drought in South Korea, *Sci. Total Environ.*, 852, 158451, <https://doi.org/10.1016/j.scitotenv.2022.158451>, 2022.
- Wells, K. C., Millet, D. B., Payne, V. H., Deventer, M. J., Bates, K. H., de Gouw, J. A., Graus, M., Warneke, C., Wisthaler, A.,

- and Fuentes, J. D.: Satellite isoprene retrievals constrain emissions and atmospheric oxidation, *Nature*, 585, 225-233, <https://doi.org/10.1038/s41586-020-2664-3>, 2020.
- 745 Wolfe, G., Kaiser, J., Hanisco, T., Keutsch, F., De Gouw, J., Gilman, J., Graus, M., Hatch, C., Holloway, J., and Horowitz, L.: Formaldehyde production from isoprene oxidation across NO_x regimes, *Atmos. Chem. Phys.*, 16, 2597-2610, <https://doi.org/10.5194/acp-16-2597-2016>, 2016.
- Woo, J.-H., Kim, Y., Kim, H.-K., Choi, K.-C., Eum, J.-H., Lee, J.-B., Lim, J.-H., Kim, J., and Seong, M.: Development of the CREATE inventory in support of integrated climate and air quality modeling for Asia, *Sustainability*, 12, 7930, <https://doi.org/10.3390/su12197930>, 2020.
- 750 Yu, J., Han, K. M., Song, C. H., Lee, K., Lee, S., Kim, Y., Woo, J.-H., Kim, S., and Wisthaler, A.: Evaluation of biogenic emissions from three different vegetation distributions in South Korea, *Atmos. Environ.*, 296, 119588, <https://doi.org/10.1016/j.atmosenv.2023.119588>, 2023.
- Zhang, X., Duan, J., Cherubini, F., and Ma, Z.: A global daily evapotranspiration deficit index dataset for quantifying drought severity from 1979 to 2022, *Sci. Data*, 10, 824, <https://doi.org/10.1038/s41597-023-02756-1>, 2023.
- 755 Zhang, X., Duan, Y., Duan, J., Jian, D., and Ma, Z.: A daily drought index based on evapotranspiration and its application in regional drought analyses, *Sci. China Earth Sci.*, 1-20, <https://doi.org/10.1007/s11430-021-9822-y>, 2022.
- Zhou, S., Medlyn, B., Sabaté, S., Sperlich, D., and Prentice, I. C.: Short-term water stress impacts on stomatal, mesophyll and biochemical limitations to photosynthesis differ consistently among tree species from contrasting climates, *Tree Physiol.*, 34, 1035-1046, <https://doi.org/10.1093/treephys/tpu072>, 2014.
- 760 Zhu, L., González Abad, G., Nowlan, C. R., Chan Miller, C., Chance, K., Apel, E. C., DiGangi, J. P., Fried, A., Hanisco, T. F., and Hornbrook, R. S.: Validation of satellite formaldehyde (HCHO) retrievals using observations from 12 aircraft campaigns, *Atmos. Chem. Phys.*, 20, 12329-12345, <https://doi.org/10.5194/acp-20-12329-2020>, 2020.
- Zscheischler, J., Martius, O., Westra, S., Bevacqua, E., Raymond, C., Horton, R. M., van den Hurk, B., AghaKouchak, A., Jézéquel, A., and Mahecha, M. D.: A typology of compound weather and climate events, *Nat. Rev. Earth Environ.*, 1, 333-347, <https://doi.org/10.1038/s43017-020-0060-z>, 2020.