Global high-resolution simulations of tropospheric nitrogen dioxide using CHASER V4.0

Takashi Sekiya¹, Kazuyuki Miyazaki¹², Koji Ogochi¹, Kengo Sudo³¹, and Masayuki Takigawa¹

¹Japan Agency for Marine-Earth Science and Technology, Yokohama, Japan
²Jet Propulsion Laboratory, California Institute of Technology, Pasadena, CA, USA
³Graduate School of Environmental Studies, Nagoya University, Nagoya, Japan

Correspondence: Takashi Sekiya (tsekiya@jamstec.go.jp)

Received: 17 August 2017 – Discussion started: 9 October 2017
Revised: 18 January 2018 – Accepted: 3 February 2018 – Published: 16 March 2018

Abstract. We evaluate global tropospheric nitrogen dioxide (NO₂) simulations using the CHASER V4.0 global chemical transport model (CTM) at horizontal resolutions of 0.56, 1.1, and 2.8°. Model evaluation was conducted using satellite tropospheric NO₂ retrievals from the Ozone Monitoring Instrument (OMI) and the Global Ozone Monitoring Experiment-2 (GOME-2) and aircraft observations from the 2014 Front Range Air Pollution and Photochemistry Experiment (FRAPPÉ). Agreement against satellite retrievals improved greatly at 1.1 and 0.56° resolutions (compared to 2.8° resolution) over polluted and biomass burning regions. The 1.1° simulation generally captured the regional distribution of the tropospheric NO₂ column well, whereas 0.56° resolution was necessary to improve the model performance over areas with strong local sources, with mean bias reductions of 67 % over Beijing and 73 % over San Francisco in summer. Validation using aircraft observations indicated that high-resolution simulations reduced negative NO₂ biases below 700 hPa over the Denver metropolitan area. These improvements in high-resolution simulations were attributable to (1) closer spatial representativeness between simulations and observations and (2) better representation of large-scale concentration fields (i.e., at 2.8°) through the consideration of small-scale processes. Model evaluations conducted at 0.5 and 2.8° bin grids indicated that the contributions of both these processes were comparable over most polluted regions, whereas the latter effect (2) made a larger contribution over eastern China and biomass burning areas. The evaluations presented in this paper demonstrate the potential of using a high-resolution global CTM for studying megacity-scale air pollutants across the entire globe, potentially also contributing to global satellite retrievals and chemical data assimilation.

1 Introduction

Nitrogen oxides (NOₓ = NO + NO₂) play a key role in air quality, tropospheric chemistry, ecosystem, and climate change. NOₓ is one of the main precursors of tropospheric ozone, a major pollutant and greenhouse gas (IPCC, 2013). Oxidation products from NOₓ, including nitric acid (HNO₃), alkyl nitrates (RONO₂), and peroxy nitrates (RO₂NO₂), are partitioned to particulate nitrates, which cause respiratory problems, degrade visibility, and affect the radiative budget by scattering solar radiation. The wet and dry deposition of nitrogen compounds affects the productivities and diversities of terrestrial and marine ecosystems on a global scale (e.g., Gruber and Galloway, 2008; Duce et al., 2008). Increasing NOₓ also reduces quantities of long-lived greenhouse gases, such as methane, due to chemical destruction via hydroxyl radicals (OH) through O₃–HOₓ–NOₓ chemistry (e.g., Shindell et al., 2009).

Major anthropogenic sources of NOₓ are ground transport and power generation, with these accounting for more than half of total global anthropogenic emissions (Janssens-Maenhout et al., 2015). NOₓ is also emitted from natural sources: biomass burning, microbial activity in soil, and lightning. Main NOₓ sinks are oxidation with OH during daytime and the hydrolysis of dinitrogen pentoxide (N₂O₅).
on aerosols during nighttime (Platt et al., 1984; Dentener and Crutzen, 1993; Evans and Jacob, 2005; Brown et al., 2006). The lifetime of NO₂, which is a function of OH concentration and NO₂ photolysis during daytime (Prather and Ehhalt, 2001), is of the order of hours to days. It also depends on aerosol surface area and composition during nighttime (e.g., Brown et al., 2006). Because of this short lifetime and heterogeneous source distribution, tropospheric NO₂ is highly variable in space and time over the globe.

Satellite observations of tropospheric NO₂ columns from the Global Ozone Monitoring Experiment (GOME), the SCanning Imaging Absorption SpectroMeter for Atmospheric CHartographY (SCIAMACHY), the Ozone Monitoring Instrument (OMI) (e.g., Duncan et al., 2016; Krotkov et al., 2016), and GOME-2 (e.g., Valsk et al., 2011; Zien et al., 2014) have been used for evaluations of chemical transport models (CTMs) (e.g., Kim et al., 2009; Huijnen et al., 2010a; Miyazaki et al., 2012; Yamaji et al., 2014). Previous model validation studies have revealed a general underestimation of simulated tropospheric NO₂ columns over polluted areas in global CTMs (van Noije et al., 2006; Huijnen et al., 2010a, b; Miyazaki et al., 2012). Global CTMs typically have a horizontal resolution of 2–5°. Meanwhile, high-resolution simulations have been conducted using regional models, which have shown the ability to simulate observed high tropospheric NO₂ columns over major polluted regions such as East Asia, North America, and Europe (e.g., Uno et al., 2007; Kim et al., 2009; Huijnen et al., 2010a; Itahashi et al., 2014; Yamaji et al., 2014; Canty et al., 2015; Han et al., 2015; Harkey et al., 2015). High-resolution simulations can lead to improvements in two ways: (1) through reduced spatial representation gaps between observed and simulated fields and (2) via improved representation of large-scale concentration fields through a consideration of small-scale processes. Using a regional CTM, Valin et al. (2011) suggested that insufficient model resolution leads to enhanced OH, shortened NO₂ lifetime, and too-low NO₂ over strong local emissions. The authors also suggested that 4 and 12 km resolutions are sufficient to accurately simulate the nonlinear effects of O₃–HO₂–NOₓ chemistry on NO₂ lifetime over power plants in Four Corners–San Juan and Los Angeles. Yamaji et al. (2014) estimated up to 60% error reduction in simulated tropospheric NO₂ columns at 20 km resolution over East Asia compared to 80 km resolution. Using a global CTM, Wild and Prather (2006) reported that NO₂ lifetime over East Asia increased by 22% when increasing model resolution from 5.6° × 5.6° to 1.1° × 1.1°. Williams et al. (2017) conducted a comprehensive evaluation of NO₂, SO₂, and CH₂O simulated in TM5-MP, revealing that increasing horizontal model resolution from 3° × 2° to 1° × 1° reduced negative NO₂ bias by up to 99% against 35% of the surface measurement sites (33 stations) in Europe. Horizontal model resolution could also be a crucial factor even for biomass burning areas because of highly varying emission sources and nonlinear chemical processes. However, previous studies have mostly focused on urban regions. Further investigations are required for both urban and biomass burning regions. Vertical model resolution could also be important through, for instance, vertical mixing between planetary boundary layers and the free troposphere (e.g., Menut et al., 2013).

Simulated global NO₂ fields provide important information on satellite retrieval and data assimilation, as well as contributing to a better understanding of the atmospheric environment (e.g., Boersma et al., 2011; Valsk et al., 2011; Miyazaki et al., 2012; Williams et al., 2017). The quality of a priori fields is important for retrieval of the tropospheric NO₂ column (Russell et al., 2011). For instance, low-resolution global CTMs poorly represent NO₂ variations across urban and rural regions, degrading the spatial variation of retrieved concentrations at high resolution. Several retrieval studies (Heckel et al., 2011; Russell et al., 2011; Lin et al., 2014) have employed high-resolution a priori fields from regional CTM simulations. The authors demonstrated improvements in these regional retrievals using high-resolution a priori fields in comparison to the ARCTRAS aircraft observation and ground-based remote sensing MAX-DOAS through, for instance, clearer separation of NO₂ profiles between urban, rural, and ocean regions and improved representations of altitude-dependent sensitivities (i.e., averaging kernels).

Global chemical data assimilation (e.g., Inness et al., 2015; Miyazaki et al., 2015) and emission inversion (e.g., Stavrakou et al., 2013; Miyazaki et al., 2017) also benefit from high-resolution global CTMs through improvements in model performance (e.g., Arellano Jr. et al., 2007) and reduced spatial representation gaps between observed and simulated fields. Several previous studies (Mijling and van der A, 2012; Ding et al., 2017b; Liu et al., 2017) demonstrated the importance of high-resolution modeling in detecting small-scale NO₂ emission sources such as urban, new power plant, and ship emissions. A systematic evaluation of high-resolution models enables us to discuss the application potentials of global high-resolution models to satellite retrievals and data assimilation.

In this study, we conduct a systematic evaluation of global high-resolution simulations of tropospheric NO₂ and related chemistry using CHASER V4.0. We focus on the impacts of horizontal model resolution on global tropospheric NO₂ simulations. Three horizontal resolutions of 2.8, 1.1, and 0.56° are evaluated using satellite and aircraft measurements. The remainder of this paper is structured as follows. Section 2 describes the model configuration and simulation settings and optimizes the simulated meteorological fields at different horizontal resolutions. Section 2.2 describes the observations used for validations. Section 4 presents the model evaluation results of tropospheric NO₂ using satellite-derived retrievals for the year 2008 and aircraft campaign observations from the 2014 Front Range Air Pollution and Photochemistry Experiment (FRAPPE). Section 5 discusses the resolution dependence of tropospheric chemistry. In Sect. 6, we then discuss the implications of this evaluation and the po-
tential benefits of applying global high-resolution CTMs. Finally, Sect. 7 provides concluding remarks.

2 Methodology

2.1 CHASER V4.0 model and simulations

CHASER V4.0 (Sudo et al., 2002; Sudo and Akimoto, 2007; Sekiya and Sudo, 2014) is a global chemical transport model developed in the framework of the MIROC-ESM Earth system model (Watanabe et al., 2011), which is coupled on-line with the MIROC-AGCM atmospheric general circulation model (AGCM) (K-1 model developers, 2004) and the SPRINTARS aerosol transport model (Takemura et al., 2005, 2009). Several updates were made from CHASER V3.0 (Sudo et al., 2002) to CHASER V4.0, which includes the consideration of aerosol species (sulfate, nitrate, ammonium, black and organic carbon, soil dust, and sea salt) and the implementation of related chemistry, radiation, and cloud processes. AGCM was also updated from the NIES/CCSR AGCM 5.7b to the MIROC-AGCM. Detailed information on the AGCM updates are provided by K-1 model developers (2004).

CHASER calculates gaseous, aqueous, and heterogeneous chemical reactions (93 species and 263 reactions), including the 

\[
\text{O}_3 - \text{HO}_x - \text{NO}_x - \text{CH}_2 - \text{CO}
\]

system with the oxidation of non-methane volatile organic compounds (NMVOCs). Major chemical reactions related to \(\text{NO}_x\) are considered, including (1) the photochemical cycle of NO and \(\text{NO}_2\), (2) oxidation of \(\text{NO}_2\) with OH, (3) heterogeneous hydrolysis of \(\text{N}_2\text{O}_5\), (4) the formation, thermal decomposition, and photolysis of peroxyacetyl nitrates (PANs), and (5) the formation of isoprene nitrates. CHASER also calculates stratospheric \(\text{O}_3\) chemistry including Chapman mechanisms and catalytic reactions related to \(\text{HO}_x\), \(\text{NO}_x\), \(\text{ClO}_x\), and \(\text{BrO}_x\) below 50 hPa. Above 50 hPa, prescribed concentrations of \(\text{O}_3\), nitrogen, and halogen species are used. Monthly ozone climatology is obtained from UGAMP (Li and Shine, 1995), whereas monthly climatologies of nitrogen and halogen species are taken from the Chemistry–Climate Model Initiative (CCMI) REF-C1SD simulation using NIES CCM (Akiyoshi et al., 2009, 2016; Morgenstern et al., 2016). Dry and wet (rain out and washout) deposition is calculated based on the resistance-based parameterization (Wesely, 1989) and cumulus convection and large-scale condensation parameterizations, respectively. Advection tracer transport is calculated using the piecewise parabolic method (Colella and Woodward, 1984) and the flux-form semi-Lagrangian scheme (Lin and Rood, 1996). The model also incorporates tracer transport on a sub-grid scale in the framework of the prognostic Arakawa–Schubert cumulus convection scheme (Emori et al., 2001) and the vertical diffusion scheme (Mellor and Yamada, 1974).

We evaluated two 1-year global simulations for tropospheric \(\text{NO}_2\) in 2008 and 2014 with a 1-year spin-up calculation for each simulation. In each case, three model calculations were conducted at different horizontal resolutions: T42 (i.e., 2.8° × 2.8°; hereinafter referred to as the 2.8° simulation), T106 (i.e., 1.1° × 1.1°; hereinafter the 1.1° simulation), and T213 (i.e., 0.56° × 0.56°; hereinafter the 0.56° simulation); 32 vertical layers from the surface to approximately 40 km altitude were used across the three simulations. To meet the Courant–Friedrich–Levy (CFL) condition, different maximal time steps were used for each resolution: i.e., 20 min at 2.8° resolution, 8 min at 1.1° resolution, and 4 min at 0.56° resolution. Sea-surface temperatures (SSTs) and sea-ice concentrations (SICs) were prescribed by HadISST for the corresponding year (Rayner et al., 2003). Simulated air temperature and horizontal wind were nudged to 12-hourly ERA-Interim reanalysis data (Dee et al., 2011). ERA-Interim reanalysis data at 0.75° × 0.75° horizontal resolution with 37 pressure levels were linearly interpolated to each model grid, possibly degrading simulated meteorological fields at finer resolution (i.e., 0.56°). We specified 5 and 0.7 days of nudging time for temperature and horizontal wind, respectively.

\(\text{NO}_x\) emissions from anthropogenic, biomass burning, lightning, and soil sources were considered. Anthropogenic emissions from the HTAP_v2.2 inventory for the year 2008 (Janssens-Maenhout et al., 2015) were employed for the 2008 simulations, with these originally having 0.1° × 0.1° resolution. For the 2014 simulation, anthropogenic emissions for the latest available year 2010 of the HTAP_v2.2 inventory were used. Biomass burning emissions were taken from the Global Fire Emissions Database (GFED) version 4.1 (0.25° × 0.25° resolution) (Giglio et al., 2013) for the two study years. Soil emissions were obtained from the Global Emission InitiAtive (GEIA) database (1° × 1°) (Yienger and Levy, 1995). Model of Emissions of Gases and Aerosols from Nature (MEGAN) version 2 data (0.5° × 0.5°) were used for biogenic NMVOCs emissions (Guenther et al., 2006). Annual mean total global \(\text{NO}_x\) emissions from the surface were 45.3 and 45.9 Tg N yr\(^{-1}\) in 2008 and 2014, respectively. Lightning \(\text{NO}_x\) sources were calculated as a function of cloud top height in the cumulus convection parameterization (prognostic Arakawa–Schubert scheme) at each time step of CHASER, following Price and Rind (1992).

We considered diurnal cycles of surface \(\text{NO}_x\) emissions following Miyazaki et al. (2012). Different diurnal cycles were assumed depending on the dominant source category of each region: anthropogenic-type diurnal cycles (with maxima in the morning and evening, with a factor of about 1.4) over Europe, eastern China, Japan, and North America; biomass-burning-type diurnal cycles (with a rapid increase in the morning and maximum midday, with a factor of about 3) over Central Africa and South America; and soil-type diurnal cycles (with maxima in the afternoon, with a factor of about 1.2) in the grasslands or sparsely vegetated areas of Australia, Sahara, and western China. Miyazaki et al. (2012)
confirmed that the application of this scheme leads to improvements in global tropospheric NO\textsubscript{2} simulations at 2.8° resolution. Improvements were commonly found in the 1.1° resolution simulation, whereas we did not evaluate the impact at 0.56° resolution. Over biomass burning regions, the emission diurnal variability applied in this study is generally similar to variability from 3-hourly GFED4.1 data, while distinct differences in relative magnitude around the GOME-2 overpass time suggest that model performance could differ in comparison to the GOME-2 retrievals when using the 3-hourly GFED4.1 data.

The CTM–AGCM online coupling framework used in this study has advantages over the off-line CTM framework driven by meteorological analysis or reanalysis data. First, the online framework is able to simulate short-term nonlinear variations in chemical and transport processes at every time step of the model (1–20 min in this study) in contrast to offline CTMs driven by meteorological data, typically with 6-hourly intervals. Second, grid-scale and sub-grid-scale transport processes (e.g., convection, turbulent mixing) are represented in a consistent manner based on AGCM physics (e.g., mass balance) at short time intervals. Third, the online framework allows for a flexible choice of CTM resolution, whereas the off-line framework requires matching (or interpolations without physically meaningful variations) between the CTM and meteorological data resolutions.

2.2 Observations

2.2.1 Satellite tropospheric NO\textsubscript{2} retrievals

We used tropospheric NO\textsubscript{2} column retrievals from OMI and GOME-2. OMI, onboard the Aura satellite, is an ultraviolet–visible nadir-scanning solar-backscatter spectrometer covering the spectral range of 270–500 nm (Levelt et al., 2006). The Aura satellite, launched in 2004, is in a Sun-synchronous polar orbit at 705 km of altitude with a local Equator crossing time of approximately 13:40 LT. The ground pixel size of OMI ranges from 13 × 24 km\textsuperscript{2} to 26 × 128 km\textsuperscript{2} depending on the satellite viewing angle. OMI tropospheric NO\textsubscript{2} column retrievals have daily global coverage. We used the DOMINO version 2.0 data product (Boersma et al., 2011) obtained from the TEMIS website (http://www.temis.nl/). Observations with cloud radiance fraction < 0.5, surface albedo < 0.3, and quality flag 0 were used. Retrievals from 2014 affected by row anomalies were screened using a quality flag.

Tropospheric NO\textsubscript{2} retrievals from GOME-2 on MetOP-A and MetOP-B were used to compare the years 2008 and 2014, respectively. GOME-2 is a nadir-scanning ultraviolet–visible spectrometer covering the spectral range of 240–790 nm. MetOp-A, launched in 2007, and MetOp-B, launched in 2013, are on a Sun-synchronous polar orbit at 817 km with a local Equator crossing time of 09:30 LT. The ground pixel size is 80 × 40 km\textsuperscript{2}. We used the TM4NO2A version 2.3 product obtained from the TEMIS simulations (Boersma et al., 2004). The GOME-2 retrievals were derived with the same basic algorithm as in DOMINO version 2 (Boersma et al., 2011).

For model–retrieval comparisons, we first sampled simulated NO\textsubscript{2} profiles at the closest times to measurement using 2-hourly model outputs; these were then linearly interpolated in time to the center of each measurement from the four surrounding model grids. Second, averaging kernels (AKs) were applied to the interpolated model profiles in order to consider the altitude-dependent sensitivity of retrievals. Third, retrieved and simulated NO\textsubscript{2} columns were averaged on 0.5 and 2.8° bin grids for model evaluation. In order to identify the drivers of model–retrieval differences and causes of NO\textsubscript{2} error reductions in high-resolution simulations, we conducted model evaluations at 0.5 and 2.8° bin grids (i.e., the model and retrieval fields were interpolated to 0.5 and 2.8° bin grids). Improved agreement in high-resolution simulations can be attributed to two factors: (1) closer spatial representativeness between simulations and satellite retrievals (up to approximately 0.5°) and (2) improvements in mean concentration fields on a large scale (i.e., at 2.8°) through the consideration of small-scale processes. The error reductions evaluated at the 0.5° bin grid should reflect both effects, whereas error reductions evaluated at the 2.8° bin grid should mainly be attributed to the latter effect (2). When error reductions evaluated at the 2.8° bin grid are about half the error reductions evaluated at the 0.5° bin grid, the contributions of the two effects should be identical. When error reductions evaluated at 2.8 and 0.5° bin grids are comparable, the latter effect (2) should be dominant.

It should be noted that tropospheric NO\textsubscript{2} retrievals from SCIAMACHY were also available for 2008. The model evaluation results are generally similar between GOME-2 and SCIAMACHY. Results using SCIAMACHY are not discussed in this paper.

2.2.2 Aircraft observation data

Vertical profiles of NO, NO\textsubscript{2}, OH, HO\textsubscript{2}, O\textsubscript{3}, H\textsubscript{2}O, and the photolysis rate of O\textsubscript{3} to O(1D) were obtained from the 2014 Front Range Air Pollution and Photochemistry Experiment (FRAPPÉ) campaign (Vu et al., 2016). The FRAPPÉ campaign was conducted using the NSF/NCAR C130 aircraft during the period from 16 July through 18 August 2014. The C130 flight track covered the northern Colorado plains and foothills and the area west of the Continental Divide. NO, NO\textsubscript{2}, and O\textsubscript{3} concentrations were measured by two-channel (for NO and NO\textsubscript{2}) and one-channel (for O\textsubscript{3}) chemiluminescence instruments (Ridley et al., 2004). OH and HO\textsubscript{2} were analyzed using a CIMS-based instrument that is part of the Mauldin–Cantrell HO\textsubscript{3} CIMS instrument (e.g., Mauldin et al., 2003). Water vapor was measured by a wavelength-scanned cavity ring-down spectroscopy (WS-CRDS) analyzer. The photolysis rate of O\textsubscript{3}
to O(1D) data, calculated from NCAR HARP–CFAS (CCD-based actinic flux spectroradiometer), were used. We used 1 min merged data obtained from the NASA LaRC Airborne Science Data for Atmospheric Composition (http://www-air.larc.nasa.gov/). For comparison purposes, we sampled simulated profiles at the closest time to measurement using 2-hourly model outputs; these were then linearly interpolated to measurement from the four surrounding model grids in the horizontal. The observed and simulated vertical profiles were compared by averaging data within each vertical pressure bin: 850 hPa (using data between the surface and 825 hPa), 800 hPa (825–775 hPa), 750 hPa (775–725 hPa), 700 hPa (725–675 hPa), 650 hPa (675–625 hPa), 600 hPa (625–575 hPa), 550 hPa (575–525 hPa), and 500 hPa (525–475 hPa).

2.2.3 Ozoneonde

Simulated vertical profiles of tropospheric ozone were also evaluated using ozonesonde observations. The observed vertical profiles of ozone were obtained from the World Ozone and Ultraviolet Data Center (Woudc, https://woudc.org/), the Southern Hemisphere ADditional OZonesondes (SHADOZ, https://tropo.gsfc.nasa.gov/shadoz/) (Thompson et al., 2003a, b), and the NOAA Earth System Research Laboratory (ESRL) Global Monitoring Division (GMD, https://www.esrl.noaa.gov/gmd/). All available data from these sources were used. The observed and simulated ozone profiles were compared at ozonesonde locations by averaging data within each vertical pressure bin: 850 hPa (875–825 hPa), 500 hPa (550–450 hPa), 300 hPa (350–275 hPa), and 100 hPa (112.5–92.5 hPa).

3 Validations of meteorological fields

In the CTM-AGCM online framework, meteorological fields vary among different model resolutions. From sensitivity calculations, the strength and distribution of the cumulus convection were found to be sensitive to model resolution. The cumulus convection parameterization for 2.8° resolution was optimized following Watanabe et al. (2011). We then attempted to optimize the relevant model parameters (critical relative humidity for cumulus convection and ice-fall speed) for 1.1 and 0.56° resolutions. The criterion was to minimize the root mean square error (RMSE) of annual global total flash against the Lightning Imaging Sensor (LIS), outgoing longwave radiation (OLR) against the NOAA 18 satellite observations (Liebmann, 1996), and precipitation against the Global Precipitation Climatology Project (GPCP) (Adler et al., 2003; Huffman et al., 2009) for the year 2008. The obtained minimum values of RMSE for annual mean flash rate were 0.010, 0.011, and 0.011 flashes km⁻² day⁻¹ at 2.8, 1.1, and 0.56° resolutions, respectively. Optimizing the cumulus convection setting reduced the positive bias of the annual global mean OLR by 80% at 1.1° resolution and by 50% at 0.56° resolution. Simulated global flash frequency and annual global lightning NO₂ sources (in brackets) varied slightly: 43 flashes s⁻¹ (5.4 Tg N yr⁻¹) at 2.8° resolution, 47 flashes s⁻¹ (5.6 Tg N yr⁻¹) at 1.1° resolution, and 46 flashes s⁻¹ (5.5 Tg N yr⁻¹) at 0.56° resolution.

We also evaluated relevant meteorological fields (i.e., precipitation and cloud) that have large impacts on chemistry simulations in the online CTM framework (e.g., Hess and Vukicevic, 2003). In comparison to the GPCP precipitation data, all simulations showed similar spatial error patterns after optimization (Fig. 1a–d), having positive biases (typically by a factor of 2) north and south of the intertropical convergence zone (ITCZ) over the Pacific, Indian subcontinent, and Central Africa and negative biases in the South Pacific convergence zone (SPCZ), west of the Maritime Continent, over the Amazon, and over the southeastern United States because of the use of the same physical package (e.g., cumulus convection scheme). Meanwhile, increasing model resolution led to large error reductions by up to 70% at 1.1 and 0.56° resolutions over the northwest Pacific and Atlantic oceans (negative biases) and over the northern part of China and the western part of the North American continent (positive biases). All simulations also showed reasonable agreement with OLR derived from the NOAA 18 satellite. The global mean positive bias was 80 and 50% lower at 1.1 and 0.56° resolutions, respectively, than at 2.8° resolution (Fig. 1e–h), suggesting improved photolysis calculations in the high-resolution simulations. Among different regions, the positive model bias at 2.8° resolution was largest over the Maritime Continent, which was reduced by 86% at 1.1° resolution and by 75% at 0.56° resolution. Over northern South America, in contrast, most of the positive biases remained at 1.1 and 0.56° resolutions. The model simulations were thus appropriately set up at all resolutions, while various features of the high-resolution framework were improved. Further validation at various spatial–temporal scales for different meteorological parameters will be helpful to evaluate detailed AGCM performance, even if it is beyond the scope of the current study.

4 Validations of tropospheric NO₂ columns and profiles

4.1 Global and regional distributions

Figure 2 compares the simulated annual mean tropospheric NO₂ column with satellite retrievals. Both OMI and GOME-2 retrievals showed high tropospheric NO₂ columns over eastern China, the United States, Europe, India, Southeast Asia, Central and South Africa, and South America. For most of these regions, observed concentrations were higher in GOME-2 than OMI, reflecting the difference in overpass time and diurnal variations in tropospheric chemistry (Boersma et al., 2008). All model simulations captured the
Figure 1. Annual mean precipitation rate (mm day$^{-1}$) from GPCP (a) and outgoing longwave radiation (OLR; W m$^{-2}$) from NOAA 18 satellite (e) for 2008. The second, third, and fourth columns show differences in precipitation (b–d) and OLR (f–h) between the observations and the model simulations at 2.8$^\circ$ (b, f), 1.1$^\circ$ (c, g), and 0.56$^\circ$ (d, h) resolutions, respectively. The observations and model results are mapped onto 2.5 and 1$^\circ$ bin grids for precipitation and OLR, respectively.

Figure 2. Annual mean tropospheric NO$_2$ column ($\times 10^{15}$ molecules cm$^{-2}$) from satellite retrievals (first column; a, e) and differences between the model simulation at 2.8$^\circ$ (second column; b, f), 1.1$^\circ$ (third column; c, g), and 0.56$^\circ$ (fourth column; d, h) resolutions and satellite retrievals from OMI (upper row; a–d) and GOME-2 (lower row; e–h) for 2008. The observed and simulated fields are mapped onto a 0.5$^\circ$ bin grid. The white square line in (a) represents the region used for the model evaluation.

observed global spatial variation well, with $r > 0.9$ in comparison to both OMI and GOME-2 for annual mean concentration fields. In terms of global averages, the 2.8$^\circ$ simulations were biased on the low side by 40% compared to OMI and by 47% compared to GOME-2. This negative global mean bias has commonly been reported using other global CTMs (van Noije et al., 2006; Huijnen et al., 2010b; Miyazaki et al., 2012). As summarized in Table 1, the negative annual global mean bias compared to OMI (GOME-2) was slightly reduced by 5% (3%) at 1.1$^\circ$ resolution and by 2% (1%) at 0.56$^\circ$ resolution compared to the 2.8$^\circ$ resolution. Global RMSE was reduced by 15% compared to OMI and GOME-2 by increasing model resolution from 2.8 to 1.1$^\circ$. The improvement when increasing resolution from 1.1 to 0.56$^\circ$ was limited.
Table 1. Comparisons of annual mean tropospheric NO$_2$ column between satellite retrievals (OMI and GOME-2) and the model simulation at 2.8, 1.1, and 0.56° resolutions. MB is the mean bias, RMSE is the root mean square error. S-Corr. signifies the spatial correlation coefficient. Units of MB and RMSE are $\times 10^{15}$ molecules cm$^{-2}$. The definition of the regions is the same as in Fig. 3.

<table>
<thead>
<tr>
<th>Region</th>
<th>Model resolution</th>
<th>OMI</th>
<th>GOME-2</th>
</tr>
</thead>
<tbody>
<tr>
<td>Global</td>
<td>2.8° × 2.8°</td>
<td>−0.25</td>
<td>0.90</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.23</td>
<td>0.93</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.24</td>
<td>0.93</td>
</tr>
<tr>
<td>E-China</td>
<td>2.8° × 2.8°</td>
<td>−1.71</td>
<td>0.80</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.17</td>
<td>0.86</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.60</td>
<td>0.91</td>
</tr>
<tr>
<td>E-USA</td>
<td>2.8° × 2.8°</td>
<td>−0.36</td>
<td>0.83</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>0.046</td>
<td>0.93</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.13</td>
<td>0.96</td>
</tr>
<tr>
<td>W-USA</td>
<td>2.8° × 2.8°</td>
<td>−0.38</td>
<td>0.66</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.33</td>
<td>0.82</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.32</td>
<td>0.91</td>
</tr>
<tr>
<td>Europe</td>
<td>2.8° × 2.8°</td>
<td>−0.53</td>
<td>0.87</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.41</td>
<td>0.89</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.60</td>
<td>0.91</td>
</tr>
<tr>
<td>India</td>
<td>2.8° × 2.8°</td>
<td>−0.28</td>
<td>0.84</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.26</td>
<td>0.91</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.27</td>
<td>0.91</td>
</tr>
<tr>
<td>Mexico</td>
<td>2.8° × 2.8°</td>
<td>−0.28</td>
<td>0.61</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.27</td>
<td>0.82</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.28</td>
<td>0.93</td>
</tr>
<tr>
<td>N-Africa</td>
<td>2.8° × 2.8°</td>
<td>−0.43</td>
<td>0.84</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.38</td>
<td>0.86</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.41</td>
<td>0.83</td>
</tr>
<tr>
<td>C-Africa</td>
<td>2.8° × 2.8°</td>
<td>−0.38</td>
<td>0.77</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.29</td>
<td>0.81</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.27</td>
<td>0.80</td>
</tr>
<tr>
<td>S-Africa</td>
<td>2.8° × 2.8°</td>
<td>−2.08</td>
<td>0.61</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−1.31</td>
<td>0.93</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−1.21</td>
<td>0.97</td>
</tr>
<tr>
<td>S-America</td>
<td>2.8° × 2.8°</td>
<td>−0.57</td>
<td>0.87</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.48</td>
<td>0.84</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.50</td>
<td>0.84</td>
</tr>
<tr>
<td>SE-Asia</td>
<td>2.8° × 2.8°</td>
<td>−0.54</td>
<td>0.68</td>
</tr>
<tr>
<td></td>
<td>1.1° × 1.1°</td>
<td>−0.52</td>
<td>0.82</td>
</tr>
<tr>
<td></td>
<td>0.56° × 0.56°</td>
<td>−0.55</td>
<td>0.84</td>
</tr>
</tbody>
</table>

Figures 3 and 4 (5 and 6) compare seasonal variations in the regional and monthly mean tropospheric NO$_2$ column (regional RMSEs) against OMI and GOME-2 in 2008 using data incorporated at a 0.5° bin grid. Because the validation results are similar for OMI and GOME-2 for most cases, the results using OMI are discussed below.

Over eastern China, negative model biases at 2.8° resolution were reduced at 1.1 and 0.56° resolutions from February to July. In December, model bias varied with model resolution: −14% at 2.8° resolution, +23% at 1.1° resolution, and −7% at 0.56° resolution. Negative annual mean bias was reduced by 90% from 2.8 to 1.1° resolution and by 64% from 2.8 to 0.56° resolution, with increasing spatial correlations.
OMI vs. CHASER

Figure 3. Monthly time series of tropospheric NO\textsubscript{2} column ($\times 10^{15}$ molecules cm\textsuperscript{-2}) averaged in E-China (110–123\textdegree E, 30–40\textdegree N), E-USA (95–71\textdegree W, 32–43\textdegree N), W-USA (125–100\textdegree W, 32–43\textdegree N), Europe (10\textdegree W–30\textdegree E, 35–60\textdegree N), India (68–88\textdegree E, 8–35\textdegree N), Mexico (115–90\textdegree W, 15–25\textdegree N), N-Africa (20\textdegree W–40\textdegree E, 0–20\textdegree N), C-Africa (10–40\textdegree E, 20\textdegree S–0), S-Africa (26–31\textdegree E, 28–23\textdegree S), S-America (70–50\textdegree W, 20\textdegree S–0), and SE-Asia (96–105\textdegree E, 10–20\textdegree N). The black dots are OMI retrievals, the red dashed line is the model simulation at 2.8\textdegree resolution, the yellow dashed-dotted line is the model simulation at 1.1\textdegree resolution, and the blue dotted line is the model simulation at 0.56\textdegree resolution. The vertical bars indicate mean OMI retrieval errors.

(from $r = 0.80$ at 2.8\textdegree resolution to $r = 0.86$ at 1.1\textdegree resolution and 0.91 at 0.56\textdegree resolution). Annual mean RMSE was also reduced by 32\% from 2.8 to 1.1\textdegree resolution and by 9\% from 1.1 to 0.56\textdegree resolution.

Over the eastern United States, negative annual mean bias was reduced by 87\% at 1.1\textdegree resolution and by 65\% at 0.56\textdegree resolution compared to 2.8\textdegree resolution. The seasonal bias reduction reached 95\% at 0.56\textdegree resolution in summer. Annual RMSE was reduced by 37\% at 1.1\textdegree resolution and by 40\% at 0.56\textdegree resolution compared to 2.8\textdegree resolution. The larger monthly RMSE at 1.1\textdegree than at 2.8\textdegree resolution during December is attributed to large positive biases over New Jersey and Ohio, although the reason for these is unclear. The spatial correlation for annual mean concentration fields increased from $r = 0.83$ at 2.8\textdegree resolution to $r = 0.93$ at 1.1\textdegree resolution and 0.96 at 0.56\textdegree resolution.
Over the western United States, negative annual mean bias was 13% lower at 1.1° resolution and 14% lower at 0.56° resolution compared to 2.8° resolution. In summer, the negative seasonal mean bias at 0.56° resolution was slightly larger, reflecting negative biases over rural areas. RMSE for annual mean fields was reduced by 20% from 2.8 to 1.1° resolution and by 23% from 1.1 to 0.56° resolution. The spatial correlation for annual mean fields increased from $r = 0.65$ at 2.8° resolution to $r = 0.82$ at 1.1° resolution and 0.91 at 0.56° resolution.

Over Europe, negative model bias for annual mean concentrations was reduced by 23% from 2.8 to 1.1° resolution, but was 46% larger at 0.56° resolution than at 1.1° resolution. Large negative bias over the Po Valley at 2.8° resolution was reduced by 13% at 1.1° resolution and further reduced by 10% from 1.1 to 0.56° resolution. In contrast, negative bias over London was larger at 0.56° resolution than at 1.1° resolution by a factor of 4, leading to larger negative regional mean bias at 0.56° resolution. Simulated planetary boundary layer (PBL) height in the 0.56° simulation was substantially higher (by 20%) than ERA-Interim over London, which may partially contribute to the large NO$_2$ bias. Annual RMSE was also lower by 16% at 1.1° resolution and by 9% at 0.56° resolution than at 2.8° resolution. The spatial correlation for...
annual mean fields increased from 0.87 at 2.8° resolution to 0.91 at 0.56° resolution.

Over India, negative model biases were smaller at 1.1 and 0.56° resolutions than at 2.8° resolution during January–May, but were larger during June–September. RMSE for annual mean fields was reduced at 1.1 and 0.56° resolutions compared to 2.8° resolution, while annual RMSE was increased substantially at 1.1° resolution compared to 2.8° resolution.

Over South Africa, the negative annual mean bias was reduced by 37% at 1.1° resolution and by 43% at 0.56° resolution compared to 2.8° resolution.

Figure 5. Same as Fig. 3, but for root mean square error (RMSE) of tropospheric NO$_2$ column in comparison to OMI.
Figure 6. Same as Fig. 3, but for RMSE of tropospheric NO\(_2\) column in comparison to GOME-2.

Reduced by 46 and 56% at 1.1 and 0.56° resolutions, respectively. The spatial correlation was 0.93 and 0.97 at 0.56° resolution in contrast to 0.61 at 2.8° resolution. Model resolution higher than 1.1° was thus important for reproducing megacity-scale air pollution over the Highveld region of South Africa, which is a complex source area of coal mining, thermal power generation, metal mining, and metallurgical industry as discussed by Duncan et al. (2016).

Over the selected biomass burning regions (South America, North Africa, Central Africa, and Southeast Asia), all model simulations showed negative biases throughout the year. In most cases, bias reduction with increasing model resolution was limited because most forest fires burn over large extents. Over South America, negative bias for the annual mean concentration was 15% lower at 1.1° resolution and 12% lower at 0.56° resolution than at 2.8° resolution. Annual RMSE was reduced by 15% at 1.1° resolution and by 12% at 0.56° resolution. The smaller spatial correlation at high resolutions reflects an increased positive bias over a major biomass burning hot spot (12° S, 50° W). Over North Africa, annual RMSE was smaller by 9% at 1.1° resolution and by 3% at 0.56° resolution (compared to 2.8° resolution), whereas changes in mean bias and spatial correlation were small. Over Central Africa, negative annual mean...
bias was reduced by 24% at 1.1° resolution and by 30% at 0.56° resolution, while RMSE increased during the biomass burning season (by 11% at 1.1° resolution and by 24% at 0.56° resolution). The increased RMSE is associated with increased positive biases around 10–20°. Over Southeast Asia, RMSE for the annual mean fields was reduced by 7% at 1.1° resolution and by 5% at 0.56° resolution compared to 2.8° resolution. The increased errors over strong biomass burning hot spots in high-resolution simulations could be a result of more pronounced influences of largely uncertain inventories for individual burning points (e.g., Castellanos et al., 2015).

Negative biases with respect to GOME-2 were larger than to OMI in all simulations over most regions. The differences suggest that all model simulations underestimated high NO2 concentrations in the morning. The underestimations could be associated with insufficient vertical model resolution for capturing thin nocturnal boundary layers and uncertainties in HOx–NOx–CO–VOCs chemistry, NO2 photolysis rates, and emission diurnal cycles. The different model biases between OMI and GOME-2 could also be attributed to the bias between these retrievals. Irie et al. (2012) concluded that the bias between these retrievals is small and insignificant for East Asia, whereas the bias between these retrievals is unclear for other regions. For the most anthropogenically polluted regions, bias reductions at 0.56° (compared to 2.8° resolution) were similarly found for OMI and GOME-2. For South America and Central Africa, reductions of the negative bias at 0.56° resolution were larger in the comparison against OMI than GOME-2 during the biomass burning season, suggesting that the high-resolution simulation improves the representation of daytime photochemistry in the presence of enhanced biomass burning emissions.

For the evaluations, we used simulated and observed concentrations interpolated to a 0.5° bin grid. To identify the main drivers of improvements in the high-resolution simulation, we conducted further comparisons using two concentration fields interpolated to 2.8 and 0.5° bin grids. The drivers consist of (1) closer spatial representativeness between observations and simulations (up to approximately 0.5° resolution) and (2) better representation of large-scale (i.e., at 2.8°) concentration fields through the consideration of small-scale processes. Error reductions at the 0.5° bin grid include the effects of both drivers. In contrast, error reductions at the 2.8° bin grid are mainly attributed to the latter effect (2). When error reductions at the 2.8° bin grid are about half those at the 0.5° bin grid, the contributions of the two effects should be identical. For annual RMSE reductions, the contributions of the two effects were almost identical over the eastern United States, the western United States, and South Africa (by up to −1.9 and −0.9 × 1015 molecules cm−2 at 0.5 and 2.8° bin grids, respectively). In contrast, over eastern China, improved representations on the large scale (2) contributed up to 90% (i.e., reductions by 1.1 and 1.0 × 1015 molecules cm−2 at 0.5 and 2.8° bin grids, respectively, at 1.1° resolution). In this region, the large contribution of the second effect reflected spatially homogeneous error reductions over Hebei and Henan provinces. Over most biomass burning areas, improved representations on the large scale (2) dominated improvements in high-resolution modeling, with RMSE reductions of up to 0.072 × 1015 molecules cm−2 for the 2.8° bin grid and 0.071 × 1015 molecules cm−2 for the 0.5° bin grid. These results imply that, even for areas with homogeneous concentration and emission fields, high-resolution modeling can have significant impacts through a better representation of large-scale fields.

### 4.2 Tropospheric NO2 over strong local sources

Figure 7 compares the detailed spatial distribution of the tropospheric NO2 column in summer, as represented by OMI measurements and model simulations over four selected polluted areas: East Asia, South Asia, the western United States, and South Africa. Over East Asia, high concentrations were observed over the North China Plain, the Yangtze River Delta, the Pearl River Delta, Seoul, and Tokyo, which could mainly be attributed to emissions from traffic (Zeng et al., 2014) and large coal-fired power plants in the North China Plain (Liu et al., 2015). The 2.8° simulation underestimated these high concentrations and overestimated low concentrations over surrounding areas, probably associated with artificial mixing at coarse model resolution. The 1.1 and 0.56° simulations reduced negative biases over central eastern China, the Pearl River Delta, Seoul, Tokyo, and the western part of Japan. Over the Yellow Sea, the East China Sea, and off the Pacific coast of Japan, the positive biases at 2.8° resolution were mostly removed at 1.1 and 0.56° resolutions. Consequently, regional RMSE was 32% lower at 0.56° resolution. In contrast, high-resolution simulations led to overestimation over Beijing and the Yangtze River Delta.

Over South Asia, high concentrations were observed over large cities such as New Delhi, Chennai, Mumbai, and Kolkata in India, over Lahore and Multan in Pakistan, and around reported coal-based thermal power plants at 24° N, 83° E and 22° N, 83° E in India (Lu and Streets, 2012; Prasad et al., 2012). The 2.8° simulation was biased on the low side by up to 50% over these areas, except westward of New Delhi, as commonly reported using another coarse-resolution model at 2.8° resolution (Sheel et al., 2010). These negative biases were reduced by up to 50% at 1.1 and 0.56° resolutions, whereas high-resolution simulations reveal excessively high concentrations over New Delhi. Over rural areas, negative biases were larger at 1.1 and 0.56° resolutions, resulting in larger regional RMSE than at 2.8° resolution (see Sect. 4.1).

Over the western United States, high concentrations were observed around Los Angeles, San Francisco, Seattle, Phoenix, Salt Lake City, Denver, and the Four Corners and San Juan power plants. Negative biases were reduced with
increasing model resolution over most of these regions. In contrast, negative biases remained at 0.56° resolution around strong local sources. Over rural areas, negative biases increased with model resolution, partly reflecting suppressed artificial dilution from strong local sources. As a result, regional RMSE was reduced by 18 and 27% at 1.1 and 0.56° resolutions compared to 2.8° resolution. Errors, for instance, in soil NO\textsubscript{x} emissions in summer (e.g., Oikawa et al., 2015; Weber et al., 2015) could contribute to underestimations over rural areas.

Over South Africa, high concentrations were observed over the Highveld region of South Africa, a complex source area, as noted in Sect. 4.1. The large negative bias (92% at 2.8° resolution) in peak concentration over a power plant region in Mpumalanga Province (29.5°E, 26.2°S) was reduced to 69% at 1.1° resolution and 53% at 0.56° resolution.
Figure 8. Scatter plots of observed and simulated tropospheric NO$_2$ column ($\times 10^{15}$ molecules cm$^{-2}$) over strong local sources in East Asia (left column; a, c) and the western United States (right column; b, d) during JJA 2008 for the OMI retrievals (upper row; a–b) and GOME-2 retrievals (lower row; c–d). The red marks are the model simulation at 2.8° resolution, the yellow marks are the model simulation at 1.1° resolution, and the blue marks are the model simulation at 0.56° resolution. The horizontal bars indicate mean retrieval errors in OMI and GOME-2. For East Asia, the results are shown for Beijing (116.38° E, 39.92° N), Tianjin (117.18° E, 39.13° N), Shanghai (121.47° E, 31.23° N), Nanjing (118.77° E, 32.05° N), Guangzhou (113.27° E, 23.13° N), Shenzhen (114.10° E, 22.55° N), Seoul (126.96° E, 37.57° N), and Tokyo (139.68° E, 35.68° N). For the western United States, the results are shown for Los Angeles (118.25° W, 34.05° N), San Francisco (122.42° W, 37.78° N), Seattle (122.33° W, 47.61° N), Salt Lake City (111.88° E, 40.75° N), Phoenix (112.07° W, 33.45° N), Denver (104.88° W, 39.76° N), and the Four Corners and San Juan power plants (108.48° W, 36.69° N). The values and mean retrieval errors are averages within a 50 km distance from each strong source, with weighting function application based on the inverse of the distance from each location.

Negative bias (75% at 2.8° resolution) in the Johannesburg–Pretoria megacity area (28° E, 25.7–26.2° S) was also reduced to 54% at 1.1° resolution and 50% at 0.56° resolution. High-resolution simulations are thus important for regions with complex and strong local sources. At the same time, the remaining negative bias at 0.56° suggests that power plant and industrial emissions are underestimated, as suggested by Miyazaki et al. (2017), or that a model resolution higher than 0.56° is essential.

Figure 8 compares simulated high NO$_2$ concentrations with satellite retrievals at selected megacities. Eight strong source points were selected from East Asia and seven points from the western United States during June–July–August (JJA). We consider the summertime to be suitable for evalu-
T. Sekiya et al.: Global high-resolution NO\textsubscript{2} simulations

Figure 9. Regional distributions of tropospheric NO\textsubscript{2} column ($\times 10^{15}$ molecules cm\textsuperscript{-2}) from satellite retrievals (first column; a, e) and differences between the model simulation at 2.8\textdegree (second column; b, f), 1.1\textdegree (third column; c, g), and 0.56\textdegree (fourth column; d, h) resolutions and satellite retrievals from OMI (upper row; a–d) and GOME-2 (lower row; e–h) over Colorado state during July–August 2014. The observed and simulated fields are mapped onto a 0.5\textdegree bin grid. The DMA area is shown by the blue square in (a).

4.3 Validations using FRAPPÉ aircraft measurements

In this section, we evaluated model performance in relation to O\textsubscript{3}–HO\textsubscript{x}–NO\textsubscript{x} chemistry over the Denver metropolitan area (DMA; defined as 39–41\textdegree N and 103–105.5\textdegree W) in the western United States using the FRAPPÉ campaign observation data and satellite retrievals from July–August 2014. Figure 9 compares the spatial distribution of the tropospheric NO\textsubscript{2} columns between simulations and satellite retrievals around the FRAPPÉ locations. OMI and GOME-2 observed high tropospheric NO\textsubscript{2} columns over the DMA at around 40\textdegree N, 105\textdegree W. All models underestimated high concentrations by about 50\% at 2.8\textdegree resolution compared to OMI, with this declining by 37\% at 1.1\textdegree resolution and by 56\% at 0.56\textdegree resolution. The negative bias over the DMA was larger for GOME-2 than OMI, suggesting larger underestimations in simulated fields during mornings compared to afternoons. Outside the DMA, negative biases increased by 16\% for OMI and by 11\% for GOME-2 at 0.56\textdegree resolution compared to 2.8\textdegree resolution. As a result, RMSE against OMI by 98\% over Seattle, by 58\% over Salt Lake City, by 83\% over Phoenix, by 44\% over Denver, and by 78\% over the Four Corners and San Juan power plants (compared to the 2.8\textdegree model). These bias reductions resulted in an improved slope number at 0.56\textdegree resolution (0.31) compared to 2.8\textdegree resolution (0.15). In this region, comparison results were generally similar between OMI and GOME-2. These validation results demonstrate the capability of the 0.56\textdegree simulation to represent high concentrations over strong local sources.
Figure 10. Vertical profiles of NO (pptv) (a), NO\(_2\) (pptv) (b), NO\(_2\) during mornings (09:00–12:00 LT) (c) and afternoons (13:00–16:00 LT) (d), OH (pptv) (e), HO\(_2\) (pptv) (f), O\(_3\) (ppbv) (g), specific humidity (g kg\(^{-1}\)) (h), the photolysis rate of O\(_3\) (s\(^{-1}\)) (i), and the OH chemical production rate (molecules cm\(^{-3}\) s\(^{-1}\)) from O\(_1\)D and H\(_2\)O (j) over the Denver metropolitan area (39–41° N and 103–105.5° W) during the FRAPPÉ period (from 16 July to 18 August 2014). The black dots represent the measurements, the red dashed line is the model simulation at 2.8° resolution, the yellow dashed-dotted line is the model simulation at 1.1° resolution, and the blue dotted line is the model simulation at 0.56° resolution. The horizontal bars represent the standard deviation of the measurements.

and GOME-2 for the entire domain area was almost constant with varying model resolution.

Figure 10 compares mean vertical profiles of trace gases and reaction rates over the DMA. Large negative biases of NO and NO\(_2\) at 2.8° resolution were mostly removed at 1.1 and 0.56° resolutions below 650 hPa (by up to 88%), except at 800 hPa during daytime (09:00–16:00 LT). All simulations revealed large negative biases at 800 hPa during mornings (09:00–12:00 LT), but the bias was greater by 30% at 2.8° resolution than at 1.1 and 0.56° resolutions. Afternoon lower tropospheric high concentrations (13:00–16:00 LT) were captured well in high-resolution simulations. Strong morning–afternoon variations in the lower troposphere were underestimated by 32% at 0.56° resolution, by
48% at 1.1° resolution, and by 62% at 2.8° resolution. The remaining large bias in the morning at 0.56° resolution could be associated with insufficient vertical model resolution to represent mixing within nocturnal thin boundary layers.

Large negative OH biases at 2.8 and 1.1° resolutions at 850 hPa were reduced by 81% at 0.56° resolution. From 800 to 750 hPa, the 1.1° simulation showed the closest agreement with observations (0.5–7%), whereas the 2.8 and 0.56° simulations underestimated OH by 7–21% and overestimated OH by up to 27%, respectively. Above 700 hPa, all simulations overestimated OH with a factor of up to 2. All simulations also underestimated HO2 by 10–32% below 650 hPa, except at 800 hPa.

OH and HO2 concentrations depend greatly on NOx concentrations through O3–HOx–NOx chemistry, as well as HOx production and OH conversion reactions to peroxo radicals (HO2 and RO2) with CO and VOCs. Figure 11a shows the probability distribution function of NO from the FRAPPE aircraft observation and the model simulations at 800 hPa over the DMA. The observation revealed a wide range of NO concentrations from 10–10,000 pptv. The 2.8° simulation overestimated the occurrence of concentrations <100 pptv and underestimated the occurrence of concentrations >100 pptv. The 1.1 and 0.56° simulations captured the observed probability distribution function, although they slightly overestimated the peak frequency concentration and underestimated the occurrence of low (<100 pptv) and high (>1000 pptv) concentrations. Figure 11b shows the OH–NO relationship used to validate O3–HOx–NOx chemistry. The observation showed OH increase with increasing NO to 350 pptv and a decrease with increasing NO from 350 pptv; all simulations captured the lower part (NO < 800 pptv) of the observed NO–OH relationship, suggesting that the model realistically simulates nonlinear O3–HOx–NOx chemistry. The lack of high NO (> 800 pptv) with low OH resulted in an overestimation of mean OH concentrations at 0.56° resolution.

Figure 11c compares the HO2–NO relationship. All simulations underestimated the occurrence of high HO2 (>25 pptv) at low NO (<100 pptv). This implies an underestimation of HOx chemical production in the simulations. We evaluated HOx production from the chemical reaction of O(D) with H2O using temperature, specific humidity, O3 photolysis rate to O(D) (J(O1D) – O(D)), and O3 concentration with the assumption of O(D) equilibrium (Fig. 10g–j). All simulations underestimated HOx production, with the underestimation being smaller by 13% at 0.56° resolution at 800 hPa. The underestimation of HOx production was primarily attributable to a negative bias in O3 by 11% and J(O1D) by 2.5% at 0.56° resolution at 800 hPa. The negative biases of O3 and J(O1D) were reduced by 39% and 58%, respectively, at 0.56° (compared to 2.8°) resolution. Biases in specific humidity also had small impacts on calculated HOx production. Positive biases of specific humidity at 2.8° resolution above 750 hPa were reduced by up to 83% at 0.56° resolution. The lack of nitrous acid (HONO) in the model could explain a component of the HOx production underestimation, especially during mornings (e.g., Kanaya et al., 2001). The underestimation of OH conversion to peroxo radicals could also explain simulated errors in OH and HO2. Griffith et al. (2016) attributed OH overprediction and HO2 underprediction in a box model simulation to underestimation of total OH reactivity (i.e., missing OH sink) over the United States.

The 2014 simulations used the anthropogenic emission inventory for the year 2010 (see Sect. 2.1). The optimized NOx emissions from an assimilation of multiple species satellite measurements (Miyazaki et al., 2017) suggest that surface NOx emissions over the DMA in July–August increased by 7% from 2010 to 2014. The temporal variation, together with large uncertainties in the emission inventories, could explain part of the negative biases of NO and NO2 at 800 hPa, which also affects OH, HO2, and O3 through nonlinear chemistry processes.
5 Tropospheric NO\textsubscript{2}-related chemistry

We analyzed the simulated global distribution of O\textsubscript{3}, OH, and NO\textsubscript{x} in the year 2008 to characterize the resolution dependence of NO\textsubscript{2}-related chemistry. Figure 12 compares zonal mean concentrations of O\textsubscript{3} and OH during JJA. O\textsubscript{3} mixing ratios in the middle to high latitudes were 10–60 % larger at 1.1 and 0.56° than at 2.8° resolution. As shown in Table 2, at 1.1 and 0.56° resolutions, negative biases against ozonesonde observations were reduced by up to 8 ppbv at 850 hPa from middle to high latitudes in both hemispheres and by up to 13 ppbv at 500 hPa in the Southern Hemisphere (SH) and Northern Hemisphere (NH) middle and high latitudes. In contrast, positive model biases in the upper troposphere and lower stratosphere (UTLS) mostly increased with model resolution by up to 46 ppbv at 300 hPa in the SH and NH high latitudes. The increased positive bias at high latitudes in the UTLS was associated with strengthened downwelling, as will be discussed below. RMSE against ozonesonde was reduced by up to 8 ppbv at 850 and 500 hPa in middle and high latitudes, except at 500 hPa in the NH high latitudes.

In the tropics and subtropics, in contrast, O\textsubscript{3} concentrations were 5–20 % lower at 1.1 and 0.56° than at 2.8° resolution, reducing positive biases against ozonesonde observations from 2.8° resolution by 15 ppbv at 850 hPa in the tropics (30°S–30°N) and by up to 15 ppbv at 300 hPa in the midlatitudes of both hemispheres. In contrast, negative biases increased by 7 ppbv at 500 hPa and by 9 ppbv at 300 hPa in the tropics. RMSE was smaller by 10 ppbv at 0.56° than at 2.8° resolution at 300 hPa in the SH midlatitudes. Substantial improvements were achieved from the tropopause to lower stratosphere (i.e., at 100 hPa) by using high-resolution simulations. Overall, RMSE with respect to the globally available ozonesondes was reduced with increasing resolution (by up to 8.1 ppbv) at 850 and 500 hPa. In contrast, at 300 hPa, RMSE increased at 0.56° (by 1.2 ppbv) and 1.1° (by 9.4 ppbv) resolutions, reflecting larger RMSE at 0.56 and 1.1° resolutions in the high latitudes of both hemispheres.

Increased concentrations in the extratropics and decreased concentrations in the tropics resulted in only small differences in the global tropospheric ozone burden: −5.4 % at 1.1° resolution and −2.3 % at 0.56° resolution (compared to 2.8°). Meanwhile, the budget terms of global tropospheric O\textsubscript{3} differ significantly between simulations. High-resolution
models simulated an enhanced stratosphere–troposphere exchange (STE) of $\text{O}_3$ ($510\text{Tg yr}^{-1}$ at 1.1° resolution and $548\text{Tg yr}^{-1}$ at 0.56° resolution in contrast to $500\text{Tg yr}^{-1}$ at 2.8° resolution) and smaller $\text{O}_3$ chemical production ($4647\text{Tg yr}^{-1}$ at 1.1° resolution and $4565\text{Tg yr}^{-1}$ at 0.56° resolution in contrast to $4809\text{Tg yr}^{-1}$ at 2.8° resolution). Less $\text{O}_3$ chemical production was attributed to decelerating $\text{HO}_2 + \text{NO},\ \text{CH}_3\text{O}_2 + \text{NO},$ and $\text{RO}_2 + \text{NO}$. The estimated global mean $\text{O}_3$ chemical lifetime was longer in high-resolution simulations (26.1 days at 1.1° resolution and 26.3 days at 0.56° resolution in contrast to 25.3 days at 2.8° resolution) because of decreased water vapor in the middle and upper troposphere. Model resolution dependence on global STE and ozone chemical production has been similarly reported by Wild and Prather (2006), Stock et al. (2014), Yan et al. (2016), and Williams et al. (2017). The latitudinal distributions of $\text{O}_3$ differences between simulations were determined by both chemical (e.g., weakened chemical ozone production in the tropics) and transport (e.g., strengthened downwelling from extratropical stratosphere and upper tropospheric poleward motions from the tropics to the extratropics) processes.

$\text{OH}$ was smaller by 5–30% at 1.1 and 0.56° than at 2.8° resolution in the tropics and subtropics during JJA, resulting in smaller global burdens of tropospheric $\text{OH}$ by 13.5% at 1.1° resolution and by 12.4% at 0.56° resolution. These changes were associated with decreased $\text{HO}_x$ chemical production (i.e., $\text{O}(^1\text{D}) + \text{H}_2\text{O} \rightarrow 2\text{OH}$) and $\text{HO}_2$ to $\text{OH}$ conversion reaction (i.e., $\text{HO}_2 + \text{NO} \rightarrow \text{OH} + \text{NO}_2$) by 5% at 1.1° and 0.56° resolutions (compared to 2.8° resolution). A large relative $\text{OH}$ increment was found over the Antarctic because weak ultraviolet radiation led to small $\text{OH}$ concentrations during a polar night.

Figure 13 compares the spatial distribution of $\text{NO}_2$ and $\text{OH}$ in the lower troposphere between model simulations. Lower tropospheric $\text{NO}_2$ partial columns were larger around strong source areas and smaller over rural and coastal areas around polluted regions at 1.1 and 0.56° resolutions, primarily resulting from suppressed artificial dilution near strong sources and chemical feedback through the $\text{O}_3$–$\text{HO}_x$–$\text{NO}_x$ system, as discussed in Sect. 4. The lower tropospheric OH partial column integrated in the lowermost five model layers (approximately below 800 hPa) was smaller at 1.1 and 0.56° resolutions over most of the continents. The differences in $\text{OH}$ and $\text{NO}_2$ exhibited similar spatial patterns over polluted and biomass burning regions: e.g., $r = 0.53$ over the western United States, $r = 0.61$ over India, and $r = 0.57$ over South America. $\text{NO}_2$ and $\text{OH}$ thus interact with each other through $\text{O}_3$–$\text{HO}_x$–$\text{NO}_x$ chemical reactions. Differences in simulated meteorological fields, such as cumulus convection, water vapor, and cloud cover, could also cause $\text{OH}$ differences.

Table 3 summarizes the chemical budget of $\text{NO}_2$ in the lowermost five model layers over eastern China, the western United States, and South America during summertime in each hemisphere. Over the selected regions, the $\text{NO}_2$ burden increased with model resolution by 33% over eastern China, by 9% over the western United States, and by 23% over South America. Over eastern China and the western United States, the conversion from $\text{NO}_2$ to HNO$_3$ with OH ($\text{P} - \text{L}(\text{NO}_3)$) dominated over the net chemical production of NO$_3$ ($\text{P} - \text{L}(\text{NO}_3)$). The estimated $\text{NO}_2$ lifetime via HNO$_3$ formation ($1 / [\text{OH}][\text{M}]$) was 8% longer at 0.56° than at 2.8° resolution. A longer $\text{NO}_2$ lifetime with increasing model resolution over East Asia is consistently reported by Wild and Prather (2006). Over the western United States, the estimated $\text{NO}_2$ lifetime was longer by 6% at 1.1° than at 2.8° resolution, whereas it was shorter by 6% at 0.56° than at 1.1° resolution. Over South America, the conversion of $\text{NO}_2$ to HNO$_3$ contributed 13–20% of the total net chemical production of NO$_3$, resulting from competition against chemical conversion to peroxy acyl nitrate (PANs) and organic nitrates. The estimated $\text{NO}_2$ lifetime via HNO$_3$ formation was
Figure 13. Global distributions of (a–c) NO\textsubscript{2} partial column \((\times 10^{15} \text{ molecules cm}^{-2})\) and (d–f) OH partial column \((\times 10^{11} \text{ molecules cm}^{-2})\) integrated in the lowermost five model layers in the model simulation at 2.8\textdegree resolution (first column) during JJA in 2008 and differences between the model simulation at 1.1\textdegree (second column) and 0.56\textdegree (third column) resolutions and the model at 2.8\textdegree resolution.

Table 3. Regional net chemical production of NO\textsubscript{x} via all reactions and HNO\textsubscript{3} formation (Tg yr\(^{-1}\)), NO\textsubscript{2} burden (Gg), NO\textsubscript{2} lifetime via HNO\textsubscript{3} formation reaction (hours) in the lowermost five model layers, and planetary boundary layer (PBL) height (m) in the model simulations and ERA-Interim. The definition of the regions is the same as in Fig. 3.

<table>
<thead>
<tr>
<th>Regions</th>
<th>Model resolution</th>
<th>P–L(NO\textsubscript{x})</th>
<th>P–L(HNO\textsubscript{3})</th>
<th>NO\textsubscript{2} burden</th>
<th>(\tau_{\text{NO}_2+\text{OH} \rightarrow \text{HNO}_3})</th>
<th>PBL height</th>
</tr>
</thead>
<tbody>
<tr>
<td>E-China (JJA)</td>
<td>2.8\textdegree \times 2.8\textdegree</td>
<td>-3.01</td>
<td>-3.03</td>
<td>1.49</td>
<td>4.34</td>
<td>847</td>
</tr>
<tr>
<td></td>
<td>1.1\textdegree \times 1.1\textdegree</td>
<td>-3.71</td>
<td>-3.75</td>
<td>1.98</td>
<td>4.65</td>
<td>898</td>
</tr>
<tr>
<td></td>
<td>0.56\textdegree \times 0.56\textdegree</td>
<td>-3.70</td>
<td>-3.72</td>
<td>1.99</td>
<td>4.69</td>
<td>730</td>
</tr>
<tr>
<td></td>
<td>ERA-Interim</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>556</td>
</tr>
<tr>
<td>W-USA (JJA)</td>
<td>2.8\textdegree \times 2.8\textdegree</td>
<td>-0.873</td>
<td>-0.793</td>
<td>0.91</td>
<td>10.0</td>
<td>1316</td>
</tr>
<tr>
<td></td>
<td>1.1\textdegree \times 1.1\textdegree</td>
<td>-0.859</td>
<td>-0.820</td>
<td>1.00</td>
<td>10.7</td>
<td>1218</td>
</tr>
<tr>
<td></td>
<td>0.56\textdegree \times 0.56\textdegree</td>
<td>-0.891</td>
<td>-0.849</td>
<td>0.97</td>
<td>9.99</td>
<td>1198</td>
</tr>
<tr>
<td></td>
<td>ERA-Interim</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>1127</td>
</tr>
<tr>
<td>S-America (DJF)</td>
<td>2.8\textdegree \times 2.8\textdegree</td>
<td>-0.159</td>
<td>-0.0312</td>
<td>0.30</td>
<td>83.4</td>
<td>677</td>
</tr>
<tr>
<td></td>
<td>1.1\textdegree \times 1.1\textdegree</td>
<td>-0.204</td>
<td>-0.0283</td>
<td>0.37</td>
<td>116</td>
<td>697</td>
</tr>
<tr>
<td></td>
<td>0.56\textdegree \times 0.56\textdegree</td>
<td>-0.189</td>
<td>-0.0320</td>
<td>0.36</td>
<td>98.8</td>
<td>659</td>
</tr>
<tr>
<td></td>
<td>ERA-Interim</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td>489</td>
</tr>
</tbody>
</table>

longer by 18\% at 0.56\textdegree than at 2.8\textdegree resolution. Over other regions, the regional NO\textsubscript{2} burden increased with model resolution, whereas changes in NO\textsubscript{2} lifetime via OH oxidation varied across locations (not shown), reflecting a nonlinear chemical system involving NO\textsubscript{x} (Valin et al., 2011).

Differences in simulated meteorological fields between simulations could also have effects on NO\textsubscript{2} and related
Figure 14. Global distributions of (a–c) NO$_2$ partial column ($\times 10^{15}$ molecules cm$^{-2}$) integrated from the sixth layer to the tropopause, (d–f) convective cloud updraft ($\times 10^3$ kg m$^{-2}$ s$^{-1}$) at 500 hPa, and (g–i) vertically integrated lightning NO$_x$ production ($\times 10^{12}$ kg N m$^{-2}$ s$^{-1}$) in the model simulation at 2.8° resolution (first column) during JJA in 2008 and differences between the model simulation at 1.1° (second column) and 0.56° (third column) resolutions and the model at 2.8° resolution.

species. Improvements in PBL height are especially expected to improve NO$_2$ simulations in the lower troposphere (e.g., Lin and McElroy, 2010). Table 3 compares regional mean PBL height over eastern China, the western United States, and South America in summer between ERA-Interim reanalysis and the model simulations. The 2.8° simulation overestimated regional mean PBL height in ERA-Interim; the positive bias was reduced at 0.56° resolution by 40% over eastern China, by 62% over the western United States, and by 9% over South America.

Figure 14 shows the spatial distribution of the NO$_2$ partial column in the free troposphere, convective cloud updraft mass flux at 500 hPa, and vertically integrated lightning NO$_x$ production. The simulated NO$_2$ partial column in the free troposphere was smaller by 17% at 1.1° resolution and by 14% at 0.56° resolution than at 2.8° resolution over the northern subtropics and midlatitudes, primarily because of smaller NO$_2$ concentrations above 400 hPa. These changes in the free tropospheric NO$_2$ were in contrast to the changes in the lower tropospheric NO$_2$, which were associated with suppressed convective cloud updraft over the continents by up to 76% at 1.1 and 0.56° resolutions over the northern subtropics and midlatitudes. In contrast, over the Maritime Continent, South America, and Central Africa, the free tropospheric
NO$_2$ column was larger at 1.1° resolution by up to 18% and at 0.56° resolution by up to 20% than at 2.8° resolution, primarily reflecting increased NO$_2$ concentration between 600 and 800 hPa. Lightning NO$_x$ production is also largely different between the simulations in the tropics. Over the tropics, although the mean convective cloud updraft was weaker at 1.1 and 0.56° resolutions than at 2.8° resolution, the high-resolution simulations revealed increased ice cloud in the upper troposphere and stronger (but less frequent) convection, thus increasing lightning NO$_x$ sources, especially over Asia. Meanwhile, given the same amount of lightning NO$_x$ production (using a commonly prescribed lightning NO$_x$ field in all the simulations), the high-resolution simulations revealed slightly less ozone chemical production (by 1%) through the representation of local highly concentrated NO$_x$ plumes in July 2008 (figure not shown).

The obtained evaluation results of multiple species and meteorological fields suggest that changes in NO$_2$ with increasing model resolution can be due to complex chemical interactions and different representations of meteorological fields. Further detailed validations of individual components would therefore be helpful to identify causal mechanisms and to further reduce uncertainty in high-resolution simulations.

6 Discussion

6.1 Other model error sources

Various factors other than horizontal model resolution can lead to errors in tropospheric NO$_2$ simulation. Insufficient vertical model resolution could introduce additional errors in vertical mixing, atmospheric transport, and subsequent chemistry processes, for instance, under stable boundary layer conditions during nighttime (Menut et al., 2013). Such errors could also cause large negative NO$_2$ biases during mornings in the lower troposphere (see Sect. 4.3). More detailed validation of diurnal variations is required using ground-based observations such as MAX-DOAS and lidar in future work.

Chemical kinetics information could also have large uncertainties. Lin et al. (2012) and Stavvakou et al. (2013) suggested that the uptake of HO$_2$ on aerosols is the most important factor but remains largely uncertain. CHASER includes simplified NO$_x$–VOC chemistry related to PANs and isoprene nitrates (Sudo et al., 2002). The incorporation of more detailed NO$_x$–VOC chemistry would also be needed to improve simulated peroxy nitrates and organic nitrates, as per Ito et al. (2007, 2009) and Fischer et al. (2014).

Surface emissions are another important error source. The total amounts of anthropogenic NO$_x$ emissions in China in 2008 differ by 27% between two (highest and lowest) bottom-up inventories: EDGAR4.2 and MEIC (Saikawa et al., 2017). Ding et al. (2017a) also discussed large diversity in emission inventories over East Asia. Biomass burning NO$_x$ emissions also differ significantly between inventories: for example, the annual mean emission is 2.293 Tg yr$^{-1}$ in GFASv1.0 in contrast to 2.700 Tg yr$^{-1}$ in GFEDv3.1 over the SH Africa, as reported by Kaiser et al. (2012).

Based on data assimilation of multiple species satellite measurements, Miyazaki et al. (2017) investigated large uncertainty in anthropogenic and fire-related emission factors and a significant underestimation of soil NO$_x$ sources in bottom-up emission inventories. Using a similar approach, Miyazaki et al. (2014) optimized lightning NO$_x$ sources and indicated that the widely used lightning parameterization based on the C-shape assumption (Price and Rind, 1992; Pickering et al., 1998) has large uncertainty. Implementing these optimized emissions could improve model performance, although optimal emissions could also be dependent on model resolution.

Representations of meteorological parameters, such as cloud optical depth, temperature, water vapor, PBL height, and relevant transport and chemical processes, are also important in tropospheric NO$_2$ simulations (Lin et al., 2012). Because we employed an AGCM–CTM online coupling system, meteorological fields are simulated explicitly at each model resolution. This could help to improve the tropospheric chemistry simulation. For instance, we found that simulated regional mean PBL height is sensitive to the choice of model resolution, with the 0.56° simulation showing closer agreements with ERA-Interim reanalysis, as discussed in Sect. 4.3. Resolving small-scale cloud distributions may lead to improved photolysis and convective transport calculations in high-resolution simulations.

Nevertheless, the AGCM meteorological fields still need to be carefully validated and improved. For instance, cumulus convection and cloud parameterization calculations were sensitive to model resolution. Although the relevant model parameters have been optimized separately for each model resolution, there are still some discrepancies against observed OLR and precipitation distributions (see Sect. 2).

6.2 Nonlinearity in model error reductions

Model performance was clearly better at 0.56 and 1.1° resolutions than at 2.8° resolution in most cases. The 0.56° simulation largely improved spatial variations over eastern China, the eastern and western United States, Mexico, and South Africa, as confirmed by large RMSE reductions, especially for megacities and regions with power plants (see Sect. 4). In most cases, the improvement was smaller from 1.1 to 0.56° resolution than from 2.8 to 1.1° resolution. Meanwhile, regional RMSEs increased at 0.56° from 1.1° resolution for some cases over Europe, India, and the selected biomass burning regions, possibly related to more pronounced errors in meteorological fields for Europe and India and in biomass burning hot spot emissions.

Comparisons to aircraft measurements showed better performance of NO$_x$ simulation at high resolutions. However,
the representation of NO variability (i.e., the probability distribution function) was insufficient even at 0.56° resolution. Further improvements could be obtained using a model with resolution finer than 0.56°. For instance, Valin et al. (2011) noted that 4 and 12 km resolutions are required for Four Corners and Los Angeles, respectively, to accurately simulate the nonlinear chemical feedback of the O$_3$--HO$_x$--NO$_x$ system. Yamaji et al. (2014) reported that errors in the simulated tropospheric NO$_2$ column at 20 km resolution did not yet approach convergence over Tokyo.

Most previous high-resolution modeling studies have used regional models to simulate NO$_2$ concentration fields at high spatial resolution, primarily focusing on urban regions, with reduced or equivalent computational costs compared to global models. Several studies have demonstrated that a better representation of the long-range transport of NO$_3$ reservoir species such as peroxyacetyl nitrate (PAN) are important on simulated NO$_2$ in the free troposphere in remote areas (e.g., Hudman et al., 2004; Fischer et al., 2010, 2014; Jiang et al., 2016). A two-way nesting between regional and coarse-resolution global models (e.g., Yan et al., 2016) is able to consider both small-scale processes inside focusing regions and long-range transport over the globe, which has an advantage over regional models. An important advantage of global high-resolution models over regional models and two-way nesting systems is the ability to simulate NO$_2$ concentration fields at high resolutions over the entire globe across urban, biomass burning, and remote regions in a consistent framework. Even over remote regions, a high-resolution simulation has the potential to improve model performance through considering the effects of nonlinear chemistry in highly concentrated NO$_3$ plumes emitted from ships and lightning (Charlton-Perez et al., 2009; Vinken et al., 2011; Gressent et al., 2016). These NO$_3$ emission sources in remote regions have significant impacts on climate and air quality (Eyring et al., 2010; Holmes et al., 2014; Banerjee et al., 2014; Finney et al., 2016). It is thus important to clarify the importance of resolving small-scale sources and plumes within a global modeling framework for a better understanding of the global atmospheric environment and chemistry–climate system.

Williams et al. (2017) also showed that the differences in NO$_2$ profiles between the TM5-MP model at 3° × 2° and 1° × 1° horizontal resolutions are within a few percentage points below 850 hPa over the Pacific in boreal spring. They also showed much larger differences with changing model resolution over Texas in autumn. Model resolution impact thus varies significantly with location and season. Further investigations using other aircraft measurements would be helpful to evaluate model performance in different cases.

High-resolution chemical transport modeling requires huge computational resources: e.g., compared to the simulation at 2.8° resolution (approximately 480 s computer time for a 1-day simulation), the computational cost increased by a factor of 67 at 0.56° resolution (approximately 32 000 s computer time) and by a factor of 14 at 1.1° resolution (approximately 6700 s computer time). High-performance computing (HPC) systems are thus essential for performing high-resolution simulations. At the same time, because the size of a 3-D array is large in the high-resolution model, computational efficiency is important: e.g., efficient data throughputs in memory transfer, network communication between multiple nodes, and file input–output. In the future, further improvements in computational efficiency will be required, together with the development of HPC systems.

6.3 Application for satellite retrieval and data assimilation

An important application of high-resolution tropospheric NO$_2$ simulations is to provide a priori profile information on satellite retrieval and chemical data assimilation (Liu et al., 2017). Here, we would like to discuss the potentials of the obtained results for these applications.

Current satellite retrievals of the tropospheric NO$_2$ column use a priori NO$_2$ profiles obtained from global model simulations at relatively coarse resolutions: from TM5 at 3° × 2° in DOMINO-2 (Boersma et al., 2011) and GEOS-Chem at 2.5° × 2° in OMNO2 (Bucsela et al., 2006; Celarier et al., 2008), whereas the TROPOMI retrieval product will employ 1° × 1° resolution simulation fields from TM5 (Williams et al., 2017). To provide high-resolution (ranging from 4 to 50 km) a priori information, several regional retrievals have employed regional models (Heckel et al., 2011; Russell et al., 2011; Lin et al., 2014), showing improvements in the retrieved fields in comparison to independent observations. High-resolution a priori fields from global CTMs are important in providing consistent global datasets.

To avoid spatial representation gaps between satellite measurements and coarse-resolution global models, superobservation techniques have been employed to produce representative data before assimilation (e.g., Miyazaki et al., 2012). The average of averaging kernels over a number of retrievals within a super-observation grid does not hold any physical meaning. This may inhibit effective improvement by assimilating over regions with varying conditions. High-resolution CTMs allow for the assimilation of satellite measurements, with reduced representation gaps without any averages.

Because of the distinct nonlinearity in chemical reactions, the high-resolution assimilation of satellite measurements considering small-scale variations in background error covariance would be essential in making the best use of observational information. High-resolution chemical data assimilation could also benefit air pollutant emission estimates (e.g., Miyazaki et al., 2014, 2017; Liu et al., 2017), especially using high-resolution measurements from future satellite missions such as TOROPOMI and geostationary satellites (e.g., Sentinel-4, GEMS, TEMPO), even when model resolution is still coarser than the measurement resolution.
through improved model processes and spatial representativeness for mega-cities as demonstrated by this study.

7 Summary and conclusions

We evaluated the performance of high-resolution global NO\textsubscript{2} simulations using CHASER based on comparisons against tropospheric NO\textsubscript{2} column retrievals from two satellite sensors, OMI and GOME-2, and aircraft observations during the FRAPPÈ aircraft campaign. Three different horizontal resolutions at 0.56, 1.1, and 2.8° were evaluated.

The high-resolution models at 1.1 and 0.56° resolutions showed substantial improvements in simulating tropospheric NO\textsubscript{2}. With increasing horizontal model resolution from 2.8 to 1.1°, negative regional mean model biases (RMSEs) for annual mean tropospheric NO\textsubscript{2} column were reduced over polluted regions: e.g., by 90% (32%) over eastern China, by 13% (20%) over the western United States, and by 37% (45%) over South Africa. RMSEs were further reduced by increasing model resolution from 1.1 to 0.56° over most of the polluted regions. We emphasize large error reductions from 1.1 to 0.56° resolutions by 23% over the western United States, by 25% over Mexico, and by 20% over South Africa. The high-resolution simulation at 0.56° was also essential to capture observed high tropospheric NO\textsubscript{2} columns over strong sources such as megacities and power plants. In comparison to OMI, increasing model resolution from 2.8 to 0.56° reduced negative biases over strong local sources by 67% over Beijing, by 47% over Seoul, by 62% over Tokyo, by 30% over Los Angeles, by 74% over San Francisco, and by 78% over the Four Corners and San Juan power plants in summer. Over biomass burning regions, model performance also improved with increasing model resolution from 2.8 to 1.1° and 0.56°. For instance, RMSE was reduced by 15% at 1.1° resolution (compared to 2.8° resolution) over South America. We attempted to distinguish between two different effects that led to improvements in high-resolution modeling: (1) closer spatial representativeness between observations and simulations (up to approximately 0.5° resolution) and (2) better representation of large-scale processes (i.e., at 2.8°) through the consideration of small-scale processes, for instance, associated with nonlinear O\textsubscript{3}–HO\textsubscript{x}–NO\textsubscript{x} chemistry. The relative contributions of these two effects were mostly identical over the eastern and western United States and South Africa, whereas the latter effect was dominant over eastern China and biomass burning regions.

The comparison to FRAPPÈ aircraft observations over the DMA indicated that the 0.56° simulation greatly reduced negative biases of NO\textsubscript{2} by up to 88% from the surface to 650 hPa, while improving the representation of morning–afternoon differences below 800 hPa (with a 50% reduction at 1.1° resolution). The high-resolution simulations also improved the probability distribution of NO concentrations ranging from 100–1000 pptv. However, all simulations failed to reproduce the observed low (< 100 pptv) and high (> 1000 pptv) NO concentrations, resulting in positive biases of mean OH through nonlinear NO–OH relationships.

Changes in NO\textsubscript{2} across model simulations were associated with different representations of the tropospheric chemical and transport system. By increasing model resolution from 2.8 to 0.56°, tropospheric ozone increased by up to 60% in middle to high latitudes during JJA, while ozone decreased by up to 20% in the tropics and subtropics. These changes mostly led to improved agreements against the global ozonesonde measurements. The high-resolution simulation also lowered OH concentrations throughout the troposphere by up to 30%. The regional NO\textsubscript{2} burden was larger at 0.56° than at 2.8° in the lower troposphere, by 33% over eastern China, by 9% over the western United States, and by 23% over South America. Changes in NO\textsubscript{2} lifetime via oxidation with OH varied between locations. These model resolution dependencies suggest that NO\textsubscript{2} and OH interact with each other through nonlinear relationships between NO and OH (i.e., O\textsubscript{3}–HO\textsubscript{x}–NO\textsubscript{x} chemistry).

In conclusion, the 1.1° simulation generally captures the regional distribution of the tropospheric NO\textsubscript{2} column well, but the 0.56° resolution is essential for the simulation of high NO\textsubscript{2} concentrations on a megacity scale. Meanwhile, for Europe, India, and the selected biomass burning regions, errors increased with model resolution from 1.1 to 0.56°, possibly related to more pronounced errors in meteorological fields over Europe and India and to more pronounced influences of largely uncertain inventories for individual burning points over the selected biomass burning regions. The computational cost largely increases at 0.56° resolution, while overall improvements were small at 0.56° resolution compared to 1.1° resolution except over megacities. Therefore, we consider the horizontal resolution of approximately 1° to be a realistic option to obtain improved overall performance of global tropospheric NO\textsubscript{2} simulations.

The developed high-resolution CTM framework will be a powerful tool when combined with future high-resolution satellite observations, providing valuable information on the atmospheric environment and related long-term changes on the megacity scale. We are developing a high-resolution global chemical data assimilation system based on an ensemble Kalman filter data assimilation technique (Miyazaki et al., 2017) and the developed high-resolution CTM. A post-petascale supercomputer, also known as a post-K computer, is being developed by Japan’s FLAGSHIP 2020 project (e.g., Miyoshi et al., 2015) and will facilitate future studies using the high-resolution global chemical data assimilation system and satellite observations from a new constellation of low Earth orbit sounders (e.g., IASI, AIRS, CrIS, TROPOMI, and Sentinel-5) and geostationary satellites (e.g., Sentinel-4, GEMS, and TEMPO).
**Code and data availability.** The source code for CHASER V4.0 is not publicly available because of license restrictions. The source code can be obtained from Kengo Sudo (kengo@nagoya-u.jp) upon request. Most of the source code is written in Fortran 77 and 90. The simulation data will be available upon request to the corresponding author. Satellite retrievals of the tropospheric NO\(_2\) column from OMI and GOME-2 were obtained from the TEMIS website (http://www.temis.nl/). Measurement data from the 2014 FRAPPÉ campaign were provided by the NASA LaRC Airborne Science Data for Atmospheric Composition (http://www-air.larc.nasa.gov/). GPCP combined precipitation data were downloaded from the NASA GSFC global precipitation analysis website (https://precip.gsfc.nasa.gov/). Ozone columns data were obtained from WOUDC (https://woudc.org/), SHADOZ (https://tropo.gsfc.nasa.gov/shadoz/), and NOAA ESRL GMD (https://www.esrl.noaa.gov/gmd). Interpolated OLR data were provided by the NOAA GSFC global precipitation analysis website (https://precip.gsfc.nasa.gov/). We would like to acknowledge the use of ozonesonde data obtained from the Tropospheric Emission Monitoring Internet Service (http://www.temis.nl/index.php). We would also like to express our thanks for the use of measurement data from the 2014 Front Range Air Pollution and Photochemistry Experiment (FRAPPÉ) campaign through the NASA LaRC Airborne Science Data for Atmospheric Composition (https://www-air.larc.nasa.gov/). The GPCP combined precipitation data were provided by the NASA Goddard Space Flight Center’s Laboratory for Atmospheres, which developed the data as a contribution to the GEWEX Global Precipitation Climatology Project. We would like to acknowledge the use of ozonesonde data obtained from the World Ozone and Ultraviolet Data Center (WOUDC), the Southern Hemisphere Additional Ozone and Methane (SHADOZ), and the NOAA Earth System Research Laboratory (ESRL) Global Monitoring Division (GMD). Interpolated OLR data were provided by the NOAA/OAR/ESRL PSD from their website (http://www.esrl.noaa.gov/psd/). The Earth Simulator was used for simulations as a “Strategic Project with Special Support” of the Japan Agency for Marine-Earth Science and Technology. Some simulations were also conducted by the K computer provided by the RIKEN Advanced Institute for Computational Science through the HPCI System Research Project (project IDs: hp150288, hp160231, hp170232). We would like to thank the two anonymous reviewers for their useful comments.

**Competing interests.** The authors declare that they have no conflict of interest.

**Acknowledgements.** This work was supported by the post-K computer project Priority Issue 4 (the advancement of meteorological and global environmental predictions utilizing observational Big Data) and by the Global Environment Research Fund (S-12) of the Ministry of the Environment (MOE). We acknowledge the use of tropospheric NO\(_2\) column data from GOME-2 and OMI sensors obtained from the Tropospheric Emission Monitoring Internet Service (http://www.temis.nl/index.php). We would also like to express our thanks for the use of measurement data from the 2014 Front Range Air Pollution and Photochemistry Experiment (FRAPPÉ) campaign through the NASA LaRC Airborne Science Data for Atmospheric Composition (https://www-air.larc.nasa.gov/). The GPCP combined precipitation data were provided by the NASA Goddard Space Flight Center’s Laboratory for Atmospheres, which developed the data as a contribution to the GEWEX Global Precipitation Climatology Project. We would like to acknowledge the use of ozonesonde data obtained from the World Ozone and Ultraviolet Data Center (WOUDC), the Southern Hemisphere Additional Ozone and Methane (SHADOZ), and the NOAA Earth System Research Laboratory (ESRL) Global Monitoring Division (GMD). Interpolated OLR data were provided by the NOAA/OAR/ESRL PSD from their website (http://www.esrl.noaa.gov/psd/). The Earth Simulator was used for simulations as a “Strategic Project with Special Support” of the Japan Agency for Marine-Earth Science and Technology. Some simulations were also conducted by the K computer provided by the RIKEN Advanced Institute for Computational Science through the HPCI System Research Project (project IDs: hp150288, hp160231, hp170232). We would like to thank the two anonymous reviewers for their useful comments.

Edited by: Andrea Stenke
Reviewed by: two anonymous referees

References


T. Sekiya et al.: Global high-resolution NO2 simulations


Kim, S.-W., Heckel, A., Frost, G. J., Richter, A., Gleason, J., Burrows, J. P., McKeen, S., Hsie, E.-Y., Granier, C., and Trainer, M.: NO₂ columns in the western United States observed from space and simulated by a regional chemistry model and their im-

www.geosci-model-dev.net/11/959/2018/


T. Sekiya et al.: Global high-resolution NO2 simulations


Geosci. Model Dev., 11, 959–988, 2018


