Effects of dust aerosols on tropospheric chemistry during a typical pre-monsoon season dust storm in northern India


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Abstract. This study examines the effect of a typical pre-monsoon season dust storm on tropospheric chemistry through a case study in northern India. Dust can alter photolysis rates by scattering and absorbing solar radiation and provide surface area for heterogeneous reactions. We use the Weather Research and Forecasting model coupled with Chemistry (WRF-Chem) to simulate the dust storm that occurred during 17–22 April 2010 and investigate the contribution of different processes on mixing ratios of several key trace gases including ozone, nitrogen oxides, hydrogen oxides, methanol, acetic acid and formaldehyde. We revised the Fast Troposphere Ultraviolet Visible (F-TUV) photolysis scheme to include effects of dust aerosols on photolysis rates in a manner consistent with the calculations of aerosol optical properties for feedback to the meteorology radiation schemes. In addition, we added 12 heterogeneous reactions on the dust surface, for which 6 reactions have relative-humidity-dependent reactive uptake coefficients (γ).

The inclusion of these processes in WRF-Chem is found to reduce the difference between observed and modeled O3 from 16 ± 9 to 2 ± 8 ppbv and that in NOy from 2129 ± 1425 to 372 ± 1225 pptv compared to measurements at the high-altitude site Nainital in the central Himalayas, and reduce biases by up to 30 % in tropospheric column NO2 compared to OMI retrievals. The simulated dust storm acted as a sink for all the trace gases examined here and significantly perturbed their spatial and vertical distributions. The reductions in these gases are estimated as 5–100 %, and more than 80 % of this reduction was due to heterogeneous chemistry. The RH dependence of γ is also found to have substantial impact on the distribution of trace gases, with changes of up to 20–25 % in O3 and HO2, 50 % in H2O2 and 100 % in HNO3. A set of sensitivity analyses revealed that dust aging could change H2O2 and CH3COOH levels by up to 50 % but has a relatively small impact on other gases.

1 Introduction

Dust aerosols have gained considerable attention in the recent years not only because they constitute a major fraction of the particulate matter in the troposphere but also because they have important implications for air quality, visibility, the earth’s radiation budget (e.g., Haywood and Boucher, 2000; Seinfeld et al., 2004), biogeochemistry (e.g., Jickells et al., 2005), hydrological cycles (e.g., Miller et al., 2004; Zhao et al., 2011), and atmospheric chemistry (e.g., Dentener et al., 1996; Wang et al., 2012). The significance of dust aerosols for atmospheric chemistry has been manifested through several experimental (e.g., Goodman et al., 2000; Underwood et al., 2001; Li et al., 2006; Preszler Prince et al., 2007; Wagner et al., 2008; Cwiertny et al., 2008; Pradhan et al., 2010; Crowley et al., 2010; Chen et al., 2011; Bedjanian et al., 2013a, b) and modeling studies (e.g., Zhang et al., 1994; Dentener et al., 1996; Zhang and Carmichael, 1999; Tang et al., 2004; Martin et al., 2003; Bauer et al., 2004; Tie et al.,...
Dust aerosols can influence atmospheric chemistry by affecting the photolysis rate coefficients through interaction with incoming solar radiation and by providing surface area for heterogeneous chemistry and deposition of different trace gases. It has been suggested that mineral dust aerosols are responsible for 5–20% reduction in photolysis rate coefficients of ozone (O$_3$) and nitrogen dioxide (NO$_2$) throughout the Northern Hemisphere (e.g., Martin et al., 2003; Tie et al., 2005; Ying et al., 2011). These changes in photolysis rate coefficients then decrease the annual mean global concentration of hydroxyl radical (OH) by 9%, which in turn leads to increases in the concentrations of several volatile organic compounds (Martin et al., 2003). Heterogeneous reactions on dust surfaces generally reduce the concentration of key atmospheric trace gases such as O$_3$, nitrogen oxides, sulfur oxides and hydrogen oxides, but the amount of reported reduction in these gases varies widely, indicating that heterogeneous chemistry on dust surfaces is still not well understood. For example, the decreases in surface O$_3$ are reported as 5–40% by different studies (Zhang et al., 1994; Dentener et al., 1996; Zhang and Carmichael, 1999; Tang et al., 2004; Bauer et al., 2004; Zhu et al., 2010; Wang et al., 2012). Likewise, the decreases for sulfur dioxide (SO$_2$), nitrogen oxides and hydrogen oxides are reported as 10–50, 16–100 and 11–59% respectively.

Even though this study is focused on the impact of heterogeneous processes on dust surfaces on the tropospheric chemistry in northern India, it is worth mentioning other implications of heterogeneous chemistry here. The heterogeneous uptake of trace gases can potentially affect the physiochemical properties of dust aerosols and enhance their ability to act as cloud condensation nuclei (CCN). Dust particles are generally hydrophobic when they are emitted but become hygroscopic as they travel in the atmosphere and become coated with nitrate, sulfate and organics (e.g., Levin et al., 1996; Kelly et al., 2007; Hatch et al., 2008). The coating process will also modify the distributions of dust as well as sulfate and nitrate aerosols, which in turn will increase the scattering of solar radiation by aerosols. The increase in size of dust particles due to coating will increase the probability of their removal from the atmosphere, and such coated particles will experience less long-range transport.

Box and regional modeling studies have focused mostly on heterogeneous chemistry on East Asian and Saharan dust aerosols, whereas there have been no such studies over the northern Indian region where dust storms occur frequently during the pre-monsoon (March, April, May; MAM) season (e.g., Prasad and Singh, 2007; Hegde et al., 2007). Global modeling studies have suggested that heterogeneous chemistry on dust aerosols can reduce surface O$_3$ in northern India by 4–10% (e.g., Dentener et al., 1996; Bauer et al., 2004). However, there has been a considerable improvement in our understanding of heterogeneous chemistry on dust aerosols since these global modeling studies were conducted. A major advancement has been the demonstration of relative humidity (RH) dependence of the reactive uptake coefficient for several species such as O$_3$ (Cwiertny et al., 2008), HNO$_3$ (Liu et al., 2008), OH (Bedjanian et al., 2013a), HO$_2$ (Bedjanian et al., 2013b), SO$_2$ (Preszler Prince et al., 2007) and H$_2$O$_2$ (Pradhan et al., 2010). The renoxification process (Eq. 1), which was ignored in those global modeling studies, is also included here. In this process, the uptake of gas-phase HNO$_3$ by dust particles is followed by release of gas-phase NO$_x$ (both NO and NO$_2$) in the presence of broadband radiation (e.g., Chen et al., 2011).

$$\text{HNO}_3(g) + \text{Dust} \rightarrow \text{NO}_x(g),$$  \hspace{1cm} (1)

where (g) and (p) represent the species in gas and particle phase respectively. Therefore, it is essential to reassess the importance of dust aerosols for tropospheric chemistry in northern India by taking into account these recent advancements.

In light of the above conditions, this manuscript examines the effects of dust aerosols on the distribution of many key trace gases including O$_3$, nitrogen oxides, hydrogen oxides, methanol, acetic acid and formaldehyde by incorporating the updated information on heterogeneous reactive uptake of trace gases in MOZCART chemical mechanism of Weather Research and Forecasting model coupled with Chemistry (WRF-Chem). We also revise the Fast Troposphere Ultraviolet Visible (F-TUV) scheme to include effects of dust aerosols on photolysis rates. This extended configuration of WRF-Chem is then used to simulate the impact of a typical pre-monsoon season dust storm on the regional tropospheric chemistry in northern India. This dust storm occurred during 17–22 April 2010 in northern India. A detailed analysis of the evolution of this dust storm, dust emissions and their effect on local to regional-scale aerosol optical properties and radiation budget is presented in a companion paper (Kumar et al., 2014).

We begin by describing the WRF-Chem configuration and implementation of effect of dust aerosols on photolysis rates and heterogeneous chemistry on dust surfaces. We next briefly describe the observations used to evaluate the WRF-Chem trace gas results. In the Results section, we show the importance of dust effects on tropospheric gas chemistry by comparing the model results with observations. Sensitivity simulations are used to discuss the contributions of different processes and evaluate the uncertainties in the representation of heterogeneous chemistry. We conclude that heterogeneous chemistry mostly, partly dust-modified photolysis rates, improves the agreement between model results and observations, and that these processes lead to a reduction in O$_3$, H$_2$O$_2$, and HNO$_3$.
2 The WRF-Chem model

The Weather Research and Forecasting Model version 3.4.1 (Skamarock et al., 2008) coupled with Chemistry (Grell et al., 2005; Fast et al., 2006) is used here to simulate the distribution of trace species. The simulation domain has 120 × 90 × 51 grid points in (x, y, z) direction with a grid spacing of 30 × 30 km² extending in the vertical up to 10 hPa. The static geographical fields, such as terrain height, vegetation fraction, soil properties and land use, etc., are interpolated from 10 min (~19 km) United States Geological Survey (USGS) data to the model domain by using the WRF preprocessing system (WPS). The initial and lateral boundary conditions for the meteorological fields are obtained from the National Center for Environmental Predictions (NCEP) Final Analysis (FNL) fields available every 6 h at a spatial resolution of 1° × 1°. The resolved scale cloud physics is represented by the Thompson microphysics scheme (Thompson et al., 2004), and subgrid-scale effects of convective and shallow clouds are parameterized according to the Kain–Fritsch convective scheme (Kain, 2004). The short- and longwave radiative processes in the atmosphere are represented by the Rapid Radiative Transfer Model (RRTM) (Mlawer et al., 1997). For surface processes, the model setup uses the NOAH Land Surface model (Chen and Dudhia, 2001) and MM5 similarity scheme (Beljaars, 1994). The vertical subgrid-scale fluxes due to eddy transport in the planetary boundary layer (PBL) and the free troposphere are parameterized according to the Yonsei University (YSU) boundary layer scheme (Hong et al., 2006). Four-dimensional data assimilation (FDDA) is applied to limit model errors in the simulated meteorological fields (Lo et al., 2008).

Gas-phase chemistry is represented by the Model for Ozone and Related chemical Tracers, version 4 (MOZART-4, Emmons et al., 2010), and aerosol processes by the戈CART bulk aerosol scheme (Chin et al., 2002; Pfister et al., 2014). The GOCART model simulates five major tropospheric aerosol types including sulfate, organic carbon, black carbon, dust and sea salt, assuming externally mixed aerosols. The GOCART model does not have an aerosol thermodynamics module. The emissions of sea salt (four size bins) and dust aerosols (five size bins) are calculated online within the model. The dust emission scheme is based on Ginoux et al. (2001) and calculates size-resolved dust emissions online in five size bins ranging from 0.73 to 8.0 µm (effective radius) using the following equation:

\[
F_p = \begin{cases} 
C S \rho_p u_{10m}^2 (u_{10m} - u_t) & \text{if } u_{10m} > u_t \\
0 & \text{otherwise} 
\end{cases}
\]  

(2a)

where \( F_p \) (kg m⁻² s⁻¹) represents the emission flux for size bin \( p \), \( C \) is an empirical proportionality constant (kg m⁻³ s⁻²), \( S \) is the source function representing the fraction of alluvium available for wind erosion, \( \rho_p \) is the fraction of each size class of dust in the emission, \( u_{10m} \) (m s⁻¹) is the horizontal wind speed at 10 m above the surface and \( u_t \) is the threshold velocity (m s⁻¹) below which dust emission does not occur and is a function of particle size and density, air density and surface moisture. In principle, \( S \) should be based on observations of alluvium in the model domain, but such observations do not exist. Therefore, \( S \) is obtained by comparing elevation of each grid cell with its surrounding hydrological basin spread over an area of 10° × 10°. Thus, \( S \) does not provide the exact amount of alluvium present, but it provides information about the most probable locations of sediments. The map of \( S \) is then compared with vegetation map derived from Advanced Very High Resolution Radiometer (AVHRR) data (DeFries and Townshend, 1994), and only the bare soil surfaces are considered as possible dust sources. In WRF-Chem, the source function is provided as an input static geographical field through the WRF preprocessing system (WPS). The threshold wind velocity \( u_t \) is estimated as

\[
u_t = \begin{cases} 
A \sqrt{\rho_p - \rho_a} \phi_1 \rho_{p} & \text{if } w < 0.5,
\end{cases}
\]  

(2b)

where \( A = 6.5 \) is a dimensionless parameter, \( w \) is the surface wetness (0.001–1), \( \Phi_1 \) is the particle diameter, \( g \) is the acceleration of gravity, and \( \rho_a \) and \( \rho_p \) are the air and particle density respectively. More information about estimation of \( S \) and \( u_t \) can be obtained from Ginoux et al. (2001). The tuning factor \( C \) does not have any physical meaning. Its default value was proposed initially as \( 1 \times 10^{-9} \) kg m⁻³ s⁻² (Ginoux et al., 2001), but we have set it to \( 2.2 \times 10^{-8} \) kg m⁻³ s⁻² as the latter value led to a good agreement between model and Aerosol Robotic Network (AERONET) measured aerosol optical depth and Angström exponent at seven sites in the model domain (Kumar et al., 2014). A factor of magnitude difference in the value of \( C \) results partly from the use of analysis nudging (FDDA) in this study as nudging leads to lower wind speeds. A sensitivity experiment showed that similar dust emissions could be obtained with a \( C \) value of \( 9 \times 10^{-9} \) kg m⁻³ s⁻² without applying analysis nudging. However, we chose to use analysis nudging as the model configuration with analysis nudging is found to better reproduce the variations in observed aerosol optical properties as compared to one without analysis nudging (not shown).

Anthropogenic emissions of CO, NOX, SO2, NH3, OC and BC and non-methane volatile organic compounds (NMVOCs) are taken from the MACCity emission inventory (Granier et al., 2011) and emissions for PM2.5 and PM10 are taken from the Intercontinental Chemical Transport Experiment – Phase B (INTEX-B) inventory (Zhang et al., 2009). Daily varying emissions of trace species from biomass burning are taken from the Fire Inventory from NCAR version 1 (FINN v1) (Wiedinmyer et al., 2011) and distributed vertically in the model following the online plume-rise module (Freitas et al., 2007). Note that FINN v1 accounts only for...
open biomass burning, and the residential biomass burning is included in anthropogenic emissions. Biogenic emissions of trace species from terrestrial ecosystems are calculated on-line using the Model of Emissions of Gases and Aerosols from Nature (MEGAN) version 2.04 (Guenther et al., 2006). The aerosols are allowed to provide feedback to the radiation scheme in the simulations but through direct effects only. The simulations start on 10 April 2010 at 00:00 GMT and end on 25 April 2010 at 18:00 GMT. The first 3 days of the model output are removed from the analysis to allow the model to spin up.

2.1 Effect of aerosols on photolysis rates in WRF-Chem

In this study, we use the Fast-Troposphere Ultraviolet Visible (F-TUV) scheme to calculate photolysis rates for the MOZCART chemical mechanism. The F-TUV scheme is a simplified version of the National Center for Atmospheric Research’s (NCAR) TUV model (Madronich and Weller, 1990) and is designed to reduce the computational costs associated with TUV (Tie et al., 2005). The F-TUV model utilizes the same physical processes as the TUV model except that the number of wavelengths in the TUV spectra (121–750 nm) was reduced from 140 to 17, making it eight times faster than the TUV model. The differences in the calculated photolysis rate coefficients between the TUV and F-TUV model are reported to be less than 5 % (Tie et al., 2005).

To include the effect of aerosols on photolysis rates, the F-TUV photolysis scheme in WRF-Chem calculates optical properties (optical depth, single scattering albedo and asymmetry parameter) for black carbon, organic carbon, sulfate and sea-salt aerosols, and passes them to a two-stream radiative transfer module where they interact with radiation to affect photolysis rate coefficients. These optical properties are also calculated in the aerosol optical driver of WRF-Chem and are used for feedback of aerosols on the meteorology radiation schemes. Here, we make use of the optical driver calculated aerosol optical properties in F-TUV photolysis scheme to be consistent between the effects of aerosols on radiation and photolysis rate coefficients. This coupling automatically accounts for the effect of dust aerosols on photolysis rates in WRF-Chem, which was missing in the original F-TUV scheme. Further details regarding the coupling of the optical driver to the F-TUV photolysis scheme can be found in the supplementary material.

2.2 Heterogeneous chemistry on dust surface in WRF-Chem

This study also extends the ability of the MOZCART chemical mechanism of WRF-Chem to simulate heterogeneous chemistry on the surface of dust particles by including 12 heterogeneous reactions listed in Table 1. The uptake of these gases on dust particles is assumed to be irreversible (Zhang and Carmichael, 1999) and does not add to the aerosol mass in the model because the GOCART model does not have a thermodynamics module. However, recent laboratory experiments have shown that the uptake of HNO3 (Chen et al., 2011) is associated with release of gas-phase NOx, and uptake of OH (Bedjanian et al., 2013a) and HO2 (Bedjanian et al., 2013b) are associated with the release of gas-phase H2O2. The production of these gas-phase species from heterogeneous chemistry is taken into account in this study. The production of NOx from HNO3 uptake is observed only in the presence of broadband irradiation with a yield of about 50 % (Chen et al., 2011), and thus the reaction of HNO3 is set to yield 0.5 NO2 during daytime in this study. This yield of 0.5 results from the assumption that all HNO3 molecules reacting with dust particles adsorb to the surface, but HNO3 has been observed to be lost irreversibly when reacting with CaCO3 and adsorb to the surface when reacting with other minerals (e.g., Al2O3, Fe2O3, and TiO2) (Grassian, 2002). For Indian dust particles, the abundance of CaCO3 is estimated to be about 20 % (Peterson, 1968). If we assume that HNO3 will react with the remaining 80 % of the minerals adsorbing to the surface, the yield should be 0.4. However, our assumption of a yield of 0.5 should not affect the model results significantly here as it is shown later (Sect. 4.3) that the renoxification process leads only to a small change in NOx values. The yields for H2O2 from reactions of OH and HO2 with dust particles are reported to be 5 % (Bedjanian et al., 2013a) and 10 % (Bedjanian et al., 2013b) respectively and are set accordingly. These numbers differ from previous studies, which have assumed a 100 % conversion of HO2 and 0 % conversion of OH into H2O2 (e.g., Zhu et al., 2010; Wang et al., 2012).

The pseudo-first-order reaction rate coefficient (s−1) for the loss of a gas-phase species g due to heterogeneous uptake by dust particles is calculated following Heikes and Thompson (1983) as

\[
k_g = \frac{4\pi r_i D_g V N_i}{1 + Kn[x + 4(1 - \gamma)/3\gamma]}.
\]

where \( i = 1, 5 \) represents five dust size bins used by the GOCART model, \( r_i \) and \( N_i \) represents the effective radius (cm) and number density (particles cm−3) of particles in size bin \( i \), \( V \) is the ventilation coefficient and taken as unity, and \( D_g \) represents the gas-phase molecular diffusion coefficient of gas molecule \( g \) (cm2 s−1) and is calculated following Jacobson (2005) as

\[
D_g = \frac{5}{16\rho_a A d_g^2} \sqrt{\left(\frac{m_a + m_g}{m_g}\right) RT \frac{m_a}{2\pi}},
\]

where \( \rho_a \) is atmospheric mass density (g cm−3), \( A \) is the Avogadro number (6.022 × 1023 molecules mol−1), \( d_g \) is the collision diameter (cm) of gas molecule \( g \), \( R \) is the universal gas constant (8.31451 × 107 g cm−2 s−2 mol−1 K−1), \( T \) is the absolute temperature (K), and \( m_a \) (g mol−1) and \( m_g \) (g mol−1)
are the molecular weights of air and gas respectively. In Eq. (3), \( Kn \) is the dimensionless Knudsen number defined as the ratio of the effective mean free path of a gas molecule in air (\( \lambda \)) to the effective particle radius \( r_i \). \( \gamma \) represents a correction factor for anisotropic movement and is calculated as follows:

\[
\gamma = \frac{\left( \frac{3}{2} \right) Kn + 0.71} {Kn + 1}.
\]

WRF-Chem model simulates the mass mixing ratio (\( M_i \) in g g\(^{-1}\)) of dust particles, and the use of Eq. (3) requires the conversion of mass mixing ratios to the number concentration. This is performed using the following equation:

\[
N_i = \frac{M_i \rho_a}{\left( \frac{3}{2} \pi r_i^3 \right) \rho_p}.
\]

where \( \rho_p \) (g cm\(^{-3}\)) represents the mass density of the dust particles, which are taken as 2.5 and 2.65 g cm\(^{-3}\) to represent clay and silt in GOCART (Ginoux et al., 2001).

The reaction uptake coefficient (\( \gamma \)) is the most important parameter in the calculation of \( k_g \), but the uncertainties in \( \gamma \) are very large and have been reported as more than 3 orders of magnitudes for certain species such as O\(_3\) and HNO\(_3\) (Goodman et al., 2000; Underwood et al., 2001; Michel et al., 2002). Such large uncertainties make the choice of \( \gamma \) very difficult. Since this study is focusing on Thar Desert dust aerosols for which information on heterogeneous chemistry kinetics is not available, we use the best guess values reported for East Asian dust aerosols (Zhu et al., 2010) for the \( \gamma \) values of dry dust particles except for OH and HO\(_2\). The \( \gamma \) values for OH and HO\(_2\) are taken from Bedjanian et al. (2013a, b). The applied \( \gamma \) values for dry dust particles are shown in Table 1 and vary from 2.1 \( \times \) 10\(^{-6}\) for NO\(_2\) to 0.18 for OH. Lower and upper bounds of \( \gamma \) reported in the literature are also shown in Table 1. Further information about available measurements of \( \gamma \) for different species can be found in Zhu et al. (2010).

Many laboratory studies have also demonstrated the dependence of \( \gamma \) on relative humidity (RH), but such a RH dependence has generally been ignored in previous modeling studies (e.g., Dentener et al., 1996; Zhu et al., 2010; Wang et al., 2012). In this study, we include the RH dependence of \( \gamma \) for O\(_3\) (Cwiertny et al., 2008), HNO\(_3\) (Liu et al., 2008), OH (Bedjanian et al., 2013a), HO\(_2\) (Bedjanian et al., 2013b), H\(_2\)O\(_2\) (Pradhan et al., 2010), and SO\(_2\) (Preszler Prince et al., 2007). The variations of \( \gamma \) with relative humidity for different trace gases are shown in Fig. 1. The value of \( \gamma \) increases with RH for HNO\(_3\), SO\(_2\) and H\(_2\)O\(_2\) while it decreases with RH for O\(_3\), OH and HO\(_2\). The rate coefficients, i.e., \( k_g \) values, are estimated to be of the order of 10\(^{-3}\)–10\(^{-5}\) s\(^{-1}\) for OH, HO\(_2\), H\(_2\)O\(_2\), HNO\(_3\), NO\(_3\), N\(_2\)O\(_5\), and CH\(_3\)COOH, and of the order of 10\(^{-7}\)–10\(^{-8}\) s\(^{-1}\) for O\(_3\), NO\(_2\), SO\(_2\), CH\(_3\)OH, and CH\(_2\)O.

The aging of dust particles through heterogeneous uptake of acidic gases and organic compounds is another important process that can influence the uptake of trace gases by dust particles. However, the dependence of \( k_g \) on aging of mineral dust aerosols is complex and not well understood. For example, the reactive uptake of O\(_3\) (Usher et al., 2003) on HNO\(_3\) and SO\(_2\) processed dust particles is reported to increase or decrease depending on the mineralogy of the particle, coverage of the coatings, as well as ambient RH. Due to the complexity of processes involved and lack of information on all 12 heterogeneous reactions, the effect of atmospheric processing of dust particles is not included in this study except for a set of sensitivity simulations conducted to highlight the importance of this aging process. The treatment of dust aging for those sensitivity simulations is presented below.

### Table 1. The heterogeneous reactions and dry reactive uptake coefficients (\( \gamma_{\text{dry}} \)) used in this study are listed. \( \gamma_{\text{low}} \) and \( \gamma_{\text{high}} \) represent lower and upper bounds of reactive uptake coefficient reported in the literature. The rightmost column shows the references based on which relative humidity dependence of \( \gamma_{\text{dry}} \) is specified.

<table>
<thead>
<tr>
<th>Reaction</th>
<th>( \gamma_{\text{low}} )</th>
<th>( \gamma_{\text{dry}} )</th>
<th>( \gamma_{\text{high}} )</th>
<th>RH dependence</th>
</tr>
</thead>
<tbody>
<tr>
<td>O(_3) + Dust → P</td>
<td>1 ( \times ) 10(^{-6})</td>
<td>2.7 ( \times ) 10(^{-5})</td>
<td>1 ( \times ) 10(^{-4})</td>
<td>Cwiertny et al. (2008)</td>
</tr>
<tr>
<td>HNO(_3) + Dust → 0.5 NO(_x) + P</td>
<td>1 ( \times ) 10(^{-5})</td>
<td>2.0 ( \times ) 10(^{-3})</td>
<td>0.2</td>
<td>Liu et al. (2008)</td>
</tr>
<tr>
<td>NO(_2) + Dust → P</td>
<td>4 ( \times ) 10(^{-10})</td>
<td>2.1 ( \times ) 10(^{-6})</td>
<td>2 ( \times ) 10(^{-4})</td>
<td>–</td>
</tr>
<tr>
<td>NO(_3) + Dust → P</td>
<td>0.01</td>
<td>0.1</td>
<td>0.23</td>
<td>–</td>
</tr>
<tr>
<td>N(_2)O(_5) + Dust → P</td>
<td>0.01</td>
<td>0.03</td>
<td>0.1</td>
<td>–</td>
</tr>
<tr>
<td>OH + Dust → 0.05 H(_2)O(_2) + P</td>
<td>0.004</td>
<td>0.18</td>
<td>1</td>
<td>Bedjanian et al. (2013a)</td>
</tr>
<tr>
<td>H(_2)O(_2) + Dust → 0.1 H(_2)O(_2) + P</td>
<td>0.01</td>
<td>6.4 ( \times ) 10(^{-2})</td>
<td>1</td>
<td>Bedjanian et al. (2013b)</td>
</tr>
<tr>
<td>H(_2)O(_2) + Dust → P</td>
<td>8 ( \times ) 10(^{-4})</td>
<td>2 ( \times ) 10(^{-3})</td>
<td>0.18</td>
<td>Pradhan et al. (2010)</td>
</tr>
<tr>
<td>SO(_2) + Dust → P</td>
<td>5 ( \times ) 10(^{-7})</td>
<td>3.0 ( \times ) 10(^{-5})</td>
<td>2.6 ( \times ) 10(^{-4})</td>
<td>Preszler Prince et al. (2007)</td>
</tr>
<tr>
<td>CH(_3)COOH + Dust → P</td>
<td>2.4 ( \times ) 10(^{-4})</td>
<td>1 ( \times ) 10(^{-3})</td>
<td>2 ( \times ) 10(^{-3})</td>
<td>–</td>
</tr>
<tr>
<td>CH(_3)OH + Dust → P</td>
<td>4 ( \times ) 10(^{-6})</td>
<td>1 ( \times ) 10(^{-5})</td>
<td>1.9 ( \times ) 10(^{-4})</td>
<td>–</td>
</tr>
<tr>
<td>CH(_2)O + Dust → P</td>
<td>2.6 ( \times ) 10(^{-7})</td>
<td>1 ( \times ) 10(^{-5})</td>
<td>1.1 ( \times ) 10(^{-4})</td>
<td>–</td>
</tr>
</tbody>
</table>
Laboratory experiments have shown that reactive uptake of O$_3$, HNO$_3$, OH, HO$_2$, H$_2$O$_2$ and SO$_2$ used in this study.

2.3 Atmospheric aging/processing of dust particles

Figure 1. The relative humidity dependence of reactive uptake coefficient ($\gamma$) for O$_3$, HNO$_3$, OH, HO$_2$, H$_2$O$_2$ and SO$_2$ considered: first, the number of molecules of gas $g$ needed to completely coat one dust particle of a given size with a monolayer ($n_{\text{mol,g,i}}$) is calculated as the ratio of surface area of the dust particle to the surface area occupied by a gas molecule. Second, the total number of gas molecules of $g$ that can potentially adsorb to dust particles ($n_{\text{pot,g,i}}$) is calculated by multiplying $k_g$ (s$^{-1}$) estimated in Eq. (3) with gas concentration (molecules cm$^{-3}$) and time step (180 s in this case). The ratio $n_{\text{pot,g,i}}/n_{\text{mol,g,i}}$ then provides the number concentration increment $\Delta N_{a,g,i}$ of dust particles that could have aged (i.e., been completely coated with a monolayer of gas) during this time step. The sum of the increments $\Delta N_{a,g,i}$ due to all gases considered in coating gives the total increment in aged particles $\Delta N_{a,i}$. $\Delta N_{a,i}$ is then subtracted from the number concentration of fresh dust particles $N_{f,i}$ and added accordingly to the number concentration of aged particles $N_{a,i}$.

The reactions of HNO$_3$, NO$_2$, NO and N$_2$O$_5$ with dust particles are assumed to coat dust particles with nitrate and that of SO$_2$ is assumed to coat them with sulfate. All gases are given equal probability to react with dust particles, and in case the number concentration of fresh dust particles is limiting, then $\Delta N_{a,i} > N_{f,i}$, which would lead to negative number concentrations of $N_{f,i}$. To overcome this problem, $N_{a,i}$ and $N_{f,i}$ are set to $N_{f,i}$ and zero respectively. This approach leads to mass conservation of tracers, and the sum of fresh and aged dust particle concentrations is always equal to the total dust number concentration. Both fresh and coated dust particles then react separately with trace gases.

2.4 Simulations conducted

Eleven simulations were conducted (Table 2) to examine the impact of the dust storm on tropospheric chemistry. No_Dust serves as a base simulation in which heterogeneous chemistry on dust surfaces and the effect of dust on photolysis rates are not included. Dust_J simulation includes the effect of dust aerosols on photolysis rates, while Dust_JH simulation is same as Dust_J but with addition of heterogeneous chemistry on dust surface taking into account the RH dependence of $\gamma$ and renoxification process. Dust_JH_NoReNO$_2$ simulation is same as Dust_JH except that it does not include the RH dependence of $\gamma$. In Dust_JH_NoReNO$_2$, the release of gas-phase NO$_2$ associated with uptake of HNO$_3$ is excluded in order to assess the importance of the renoxification process. Dust_JH_LoG and Dust_JH_HiG simulations are conducted with lower and upper bounds of $\gamma$ reported in the literature in order to provide an estimate of the uncertainty in heterogeneous chemistry induced changes in tropospheric chemistry due to the uncertainty in $\gamma$. Dust_JH_NO$_3$ and Dust_JH_SO$_4$ simulations are designed to examine the influences of dust coated with nitrate and sulfate separately on the uptake of O$_3$. In these simulations, the uptake coefficient $\gamma$ for aged dust particles is reduced by 70% in Dust_JH_NO$_3$ and is increased by 33% in Dust_JH_SO$_4$ (Usher et al., 2003). Dust particles are coated with both sulfate and nitrate.
in Dust_JH_NO\textsubscript{3}-SO\textsubscript{4} to examine the combined effect of nitrate and sulfate coating on O\textsubscript{3} uptake. In this simulation, three reactions of O\textsubscript{3} are included: with fresh dust particles, dust particles coated with nitrate and dust particles coated with sulfate. The original γ value is used for reaction with fresh dust particles, while the original γ value is reduced by 70% for reaction with dust particles coated with nitrate and is increased by 33% for reaction with dust particles coated with sulfate. Dust_JH_Sat is a hypothetical simulation where we assume that the presence of a nitrate or sulfate monolayer on the dust particle would saturate the dust particles and deactivate them for further catalytic uptake of O\textsubscript{3}, OH, HO\textsubscript{2}, H\textsubscript{2}O\textsubscript{2}, CH\textsubscript{3}COOH, CH\textsubscript{3}OH and HCHO, but the coating is assumed to have no effect on the uptake of HNO\textsubscript{3}, NO\textsubscript{2}, NO\textsubscript{3}, N\textsubscript{2}O\textsubscript{5} and SO\textsubscript{2}. This comparison of Dust_JH_Sat with Dust_JH and Dust_J would provide bounds of the heterogeneous chemistry induced changes in tropospheric chemistry.

3 Observation data sets

3.1 Surface observations

This study uses surface O\textsubscript{3} and NO\textsubscript{y} (sum of nitrogen oxides) observations made at the high-altitude site Nainital (79.45° E, 29.36° N, 1958 m a.m.s.l.) located in the central Himalayas. The observation site is bounded by high-altitude (2–5 km) mountains in the north and east directions and opens to the Indo-Gangetic Plain region in the south and west directions. There are no major anthropogenic sources near Nainital, and thus the observations at this site are envisaged to be representative of northern India (Kumar et al., 2010). Further details regarding the orography, vegetation cover, meteorological and chemical characteristics of Nainital can be found elsewhere (e.g., Sagar et al., 2004; Pant et al., 2006; Kumar et al., 2010; Sarangi et al., 2014). Ozone measurements are made using a standard ultraviolet absorption-based instrument, and NO\textsubscript{y} measurements are made with a chemiluminescence-based instrument. More details about the measurement setup, operating principle, accuracy, detection limits and calibration procedure are discussed elsewhere (e.g., Kumar et al., 2010; Sarangi et al., 2014). The model results are also compared with Aerosol Robotic Network (AERONET) measurements at seven sites in the model domain. Further details of AERONET and these observations sites can be found in Kumar et al. (2014).

3.2 Ozone Monitoring Instrument (OMI)

The Ozone Monitoring Instrument (OMI), aboard NASA’s Earth Observing System (EOS) Aura satellite, measures the radiation backscattered by the earth’s atmosphere and surface over the 0.27–0.5 μm wavelength range with a spatial resolution of about 13 km x 24 km at nadir in normal operational mode. These measured radiances are used for daily global retrievals of several trace species, such as O\textsubscript{3}, NO\textsubscript{2}, BrO, SO\textsubscript{2}, CH\textsubscript{2}O and aerosols. However, we do not use OMI O\textsubscript{3} and SO\textsubscript{2} retrievals because of their low sensitivity in the lower troposphere, which is the region of interest in this study. We find that OMI SO\textsubscript{2} retrievals are very noisy as average OMI-retrieved SO\textsubscript{2} planetary boundary layer (PBL) column amount values (0.5–1.0 DU) are smaller than the reported standard deviation of the noise (1.5 DU). Here, we use Level-2 tropospheric column NO\textsubscript{2} data sets made available by KNMI (Royal Netherlands Meteorological Institute) as they provide access to the averaging kernel and a priori profiles, which are needed to make a proper comparison between model profiles and satellite retrievals (e.g., Emmons et al., 2004). More details on the algorithm used to retrieve the tropospheric column NO\textsubscript{2} abundances at KNMI are described by Bucsel et al. (2006). These OMI NO\textsubscript{2} retrievals were found to correlate well with aircraft measurements made during the INTEX-B campaign (Boersma et al., 2008) and MAX-DOAS ground-based measurements (Kramer et al., 2008) but are also suggested to be biased positively by about 0–30%, irrespective of season (e.g., Boersma et al., 2009a; Zhou et al., 2009).

To compare WRF-Chem results with OMI, the best quality OMI retrievals are used by reducing influence of clouds on OMI retrievals through selection of pixels with cloud fraction less than 0.3 and removing unreliable retrievals associated with a tropospheric column flag of greater than 0 (Boersma et al., 2009b). The nighttime pixels from OMI are also excluded for the comparison. These best quality retrievals are co-located in space and time with model output. The co-located WRF-Chem profiles are then transformed using the averaging kernel and a priori profiles used in the satellite retrievals to obtain a model profile that OMI would measure for the modeled state of the atmosphere in the absence of other errors. More details about the method of model–OMI data co-location and convolution of model profiles with averaging kernel and a priori files can be found in Kumar et al. (2012).

4 Results and discussion

4.1 Model evaluation

Dust storms in northern India are characterized by large increase in aerosol optical depth (AOD) (> 50%) and decrease in α (> 70%) (Dey et al., 2004; Prasad and Singh, 2007). Both of these features were observed during this dust storm indicating that this was a typical pre-monsoon season dust storm (see Kumar et al., 2014) and can be considered as representative of dust storms in northern India. Kumar et al. (2014) also present a detailed evaluation of simulated aerosol optical properties, and here we present a summary. The simulated AOD, Ångström exponent and single scattering albedo are compared against Aerosol Robotic Network (AERONET) measurements at seven sites in the
Table 2. Simulation designs used in this study and their purpose are shown.

<table>
<thead>
<tr>
<th>Simulation index</th>
<th>Model configuration</th>
<th>Purpose</th>
</tr>
</thead>
<tbody>
<tr>
<td>No_Dust</td>
<td>WRF-Chem without dust emissions but other aerosols affects photolysis rates</td>
<td>Serves as a base simulation</td>
</tr>
<tr>
<td>Dust_J</td>
<td>WRF-Chem with dust emissions and all aerosols affects photolysis rates</td>
<td>Comparison with No_Dust will quantify the effect of dust on photolysis rates and tropospheric chemistry</td>
</tr>
<tr>
<td>Dust_JH</td>
<td>Same as Dust_J but with heterogeneous chemistry considering RH dependence of ( \gamma ) and renoxification process included</td>
<td>Comparison with No_Dust will yield total impact of dust on tropospheric chemistry, and comparison with DUST_J will yield contribution of heterogeneous chemistry</td>
</tr>
<tr>
<td>Dust_JH_NoRH</td>
<td>Same as Dust_JH but RH dependence of ( \gamma ) is not included</td>
<td>Comparison with Dust_JH will provide information on RH dependence of heterogeneous uptake of gases by dust</td>
</tr>
<tr>
<td>Dust_JH_NoReNOx</td>
<td>Same as Dust_JH but renoxification process is not included</td>
<td>Comparison with Dust_JH will provide contribution of renoxification process in NO(_x) budget.</td>
</tr>
<tr>
<td>Dust_JH_LoG</td>
<td>Same as Dust_JH but with lower limit of ( \gamma )</td>
<td>Comparison with Dust_JH_HiG provides uncertainty in the effects of heterogeneous chemistry due to uncertainty in ( \gamma )</td>
</tr>
<tr>
<td>Dust_JH_HiG</td>
<td>Same as Dust_JH but with upper limit of ( \gamma )</td>
<td></td>
</tr>
<tr>
<td>Dust_JH_NO3</td>
<td>Same as Dust_JH but the effect of nitrate-coated dust on uptake of O(_3) is included</td>
<td>Comparison with Dust_JH will provide information on changes in O(_3) uptake due to coating of dust with nitrate</td>
</tr>
<tr>
<td>Dust_JH_SO4</td>
<td>Same as Dust_JH but the effect of sulfate-coated dust on uptake of O(_3) is included</td>
<td>Comparison with Dust_JH will provide information on changes in O(_3) uptake due to coating of dust with sulfate</td>
</tr>
<tr>
<td>Dust_JH_NO3_SO4</td>
<td>Same as Dust_JH but the effect of nitrate- and sulfate-coated dust on uptake of O(_3) is included</td>
<td>Comparison with Dust_JH will provide information on changes in O(_3) uptake due to coating of dust with nitrate and sulfate</td>
</tr>
<tr>
<td>Dust_JH_Sat</td>
<td>Same as Dust_JH but nitrate- and sulfate-coated particles are assumed to be saturated</td>
<td>Comparison with Dust_JH will provide upper limit on the effect of dust aging on the tropospheric chemistry</td>
</tr>
</tbody>
</table>

model domain. It is found that the model generally underestimates the AOD over the model domain but is able to capture the temporal variations \( (r = 0.5–0.88) \) seen in AERONET measurements. The good agreement between modeled and AERONET observed Angström exponent indicates that the model is able to capture dust-storm-induced variations in aerosol size. The comparison of model results with Moderate Resolution Imaging Spectroradiometer (MODIS) AOD retrievals shows that the model is also able to capture the spatial pattern of dust-storm-induced changes in the MODIS AOD as well as the spatial pattern of the dust plume. The average MODIS and WRF-Chem AOD (550 nm) values in the high dust-laden region are estimated as 0.80 ± 0.30 and 0.68 ± 0.28 respectively.

The variations in observed and WRF-Chem-simulated (No_Dust, Dust_J, Dust_JH_NoRH and DUST_JH) daily average O\(_3\) and NO\(_x\) at Nainital during 13–24 April 2010 are shown in Fig. 2. The modeled time series of relative humidity and dust mass concentration for particles of 0.73, 1.4 and 8.0 \( \mu \)m effective radii at Nainital are also shown. Note that the WRF-Chem model at the resolution \((30 \times 30 \text{ km}^2)\) used here is not able to capture the rapidly varying topography around Nainital, and the altitude of the site in the model is off by about 900 m. To minimize the comparison errors induced by this spatial mismatch, we first obtain the altitude...
profile of model results at Nainital by bi-linearly interpolating model output at each model level to the location of Nainital \((79.45^\circ\text{E}, 29.36^\circ\text{N})\) and then linearly interpolate the resulting altitude profile to the height of Nainital \((1958\text{ m})\). Dust mass concentrations started increasing at Nainital on 17 April 2010, reached a maximum during 20–22 April and decreased thereafter. The mass concentrations of 8.0 \(\mu\text{m}\) particles remain close to zero even during the dust storm because these particles do not travel far from the source regions due to their higher deposition velocities resulting in dry deposition.

The observed \(\text{O}_3\) mixing ratios show a decrease during the dust storm period \((17–22\text{ April 2010})\), while observed \(\text{NO}_2\) mixing ratios show an increase during 17–18 April 2010 and a decrease thereafter. Analysis of back-air trajectories arriving at Nainital (not shown) showed that the increase in \(\text{NO}_2\) levels at Nainital during 17–18 April was associated with passage of low-altitude air masses over the Indo-Gangetic Plain region before arriving at Nainital, while air masses on previous days \((15–16\text{ April})\) passed over relatively cleaner regions to the north and west at higher altitudes. The WRF-Chem model without incorporating the effects of dust aerosols \((\text{No}_\text{Dust})\) does not capture the observed decrease in \(\text{O}_3\) and \(\text{NO}_2\) levels at Nainital. Including effects of dust on photolysis rates \((\text{Dust}_J)\) induces a slight decrease \((\text{about 2}\text{ ppbv})\) in modeled \(\text{O}_3\) levels, but modeled values are still significantly higher than the observations. The introduction of heterogeneous chemistry without including RH dependence of \(\gamma\) \((\text{Dust}_\text{JH}_\text{NoRH})\) in the model leads to the observed decreases in \(\text{O}_3\) and \(\text{NO}_2\), but compared to measurements we find too strong of decreases in \(\text{O}_3\) levels and slightly higher \(\text{NO}_2\) levels than the observations at Nainital during the peak of dust storm, i.e., 20–22 April 2010. The inclusion of heterogeneous chemistry with RH dependence of \(\gamma\) \((\text{Dust}_\text{JH})\) leads to the best agreement between WRF-Chem-simulated and observed \(\text{O}_3\) and \(\text{NO}_2\) values at Nainital. The WRF-Chem-simulated average \(\text{O}_3\) values at Nainital in Dust\_JH and No\_Dust configurations during 17–22 April 2010 are estimated to be 58 ± 5 ppbv and 73 ± 6 ppbv respectively as compared to the average observed value of 56 ± 10 ppbv. The corresponding WRF-Chem \(\text{NO}_2\) average values at Nainital are estimated as 1189 ± 751 pptv and 2945 ± 876 pptv respectively, as compared to the average observed value of 843 ± 887 pptv. Thus, including the effects of dust aerosols in WRF-Chem reduces the difference between average modeled and observed \(\text{O}_3\) from 16 ± 9 to 2 ± 8 ppbv and that in \(\text{NO}_2\) from 2102 ± 1425 to 346 ± 1225 pptv respectively.

The spatial distributions of average OMI-retrieved and WRF-Chem-simulated \((\text{Dust}_\text{JH}\text{ and No}_\text{Dust})\) tropospheric column \(\text{NO}_2\) during 13–16 April 2010 and 17–22 April 2010 are shown in Fig. 3. The percentage differences in tropospheric column \(\text{NO}_2\) between high and low dust emission periods are also shown for both OMI and WRF-Chem. Both the model and satellite data show similar spatial distributions with the highest values along the Indo-Gangetic Plain region during both low and high dust emission periods, but WRF-Chem generally overestimates the OMI retrievals which is consistent with previous studies over the Indian region \((\text{Kumar et al., 2012; Ghude et al., 2013})\). However, the comparison between Dust\_JH and No\_Dust configurations of WRF-Chem shows that the inclusion of effects of dust aerosols improves the model performance and reduces the model bias with respect to OMI retrievals by up to 30% especially in the Indo-Gangetic Plain region. The remaining bias in the model could be due to uncertainties in \(\text{NO}_2\) emission estimates in this region. The percentage difference plots show that the WRF-Chem model is able to capture several features of changes in OMI-retrieved spatial distribution of tropospheric column \(\text{NO}_2\) between high and low dust emission periods. The domain-averaged OMI and WRF-Chem \((\text{Dust}_\text{JH})\) tropospheric column \(\text{NO}_2\) values over the geographical region \((70–80^\circ\text{E}, 25–30^\circ\text{E})\) of maximum dust storm impact during low dust emission period are estimated
as \((2.35 \pm 1.43) \times 10^{15}\) and \((3.95 \pm 2.43) \times 10^{15}\) molecules \(\text{cm}^{-2}\) respectively, and during high dust emission period are estimated as \((2.01 \pm 1.37) \times 10^{15}\) and \((3.41 \pm 2.80) \times 10^{15}\) molecules \(\text{cm}^{-2}\) respectively. The reduction in tropospheric column \(\text{SO}_2\) during high dust emission period in both OMI and WRF-Chem indicates that the dust storm acted as a sink for \(\text{NO}_2\).

4.2 Impact of dust storm on photolysis rate coefficients

The impact of the dust storm on photolysis rates is examined by comparing the daytime (07:30–17:30 IST or 02:00–12:00 UTC) \(\text{NO}_2\) photolysis rate coefficients calculated by the WRF-Chem model with Dust_J and No_Dust configurations (Fig. 4). \(\text{NO}_2\) photolysis rates at the surface show a strong relationship with aerosol loading and are lowest over the Indo-Gangetic Plain region, which is where the anthropogenic emissions are stronger than those over other parts of the model domain. The inclusion of dust aerosols enhances the spatial heterogeneity of \(\text{NO}_2\) photolysis rate and decreases it by 5–25\% over the Thar Desert and western Indo-Gangetic Plain region. The photolysis rate coefficients of other trace gases such as \(\text{O}_3\), \(\text{HNO}_3\), \(\text{H}_2\text{O}_2\), \(\text{CH}_2\text{O}\) and \(\text{N}_2\text{O}_5\) at the surface exhibit similar features (not shown) with decreases of the same order of magnitude. The magnitude of change in photolysis rates decreases with altitude and changes sign from negative to positive near 4 km because of the increase in actinic flux due to scattering of incoming solar radiation by dust aerosol layers underneath. The spatial structure of changes in photolysis rates at 100 hPa is similar to that at the surface with the largest increases (1–5\%) over the Thar Desert and western Indo-Gangetic Plain region.

4.3 Impact of dust storm on trace gases at the surface

To determine the impact of the dust storm on surface composition, average surface mixing ratios of \(\text{O}_3\), \(\text{SO}_2\), \(\text{NO}_x\), \(\text{HNO}_3\) and \(\text{H}_2\text{O}_2\) in the DUST_JH, which is the simulation that compared best with observations at Nainital, and the No_Dust simulation are compared for the dust event time period of 17–22 April 2010 (Fig. 5). Surface \(\text{O}_3\) shows a similar spatial distribution in both runs with the lowest values over the Arabian Sea and the highest values over the eastern Indo-Gangetic Plain region. However, the dust storm clearly leads to a reduction in \(\text{O}_3\) mixing ratios by 3–14 ppbv (5–25\%) over the Thar Desert and western Indo-Gangetic Plain region. The spatial distribution of reductions in \(\text{O}_3\) mixing ratios is consistent with the distribution of dust over the model domain, and the amount of reduction is comparable to those reported by previous studies in dust source regions (e.g., Dentener et al., 1996; Tang et al., 2004; Pozzoli et al., 2008; Wang et al., 2012).

Sulfur dioxide and \(\text{NO}_x\) mixing ratios are highest along the Indo-Gangetic Plain region due to higher emissions in this region (Fig. 5). Changes in \(\text{SO}_2\) mixing ratios show a mixed response to heterogeneous chemistry with a decrease of 0.1–0.2 ppbv (8–10\%) over the Thar Desert region and an increase of 0.2–0.5 ppbv (2–6\%) over the eastern Indo-Gangetic Plain region. This is because the heterogeneous chemistry reduces \(\text{SO}_2\) through reaction with dust particles while increases it through reduction of \(\text{OH}\) mixing ratios (\(\text{SO}_2 + \text{OH} \rightarrow \text{sulfate}\)). The sign of the changes in \(\text{SO}_2\) is thus determined by competition between these two reactions. In general, \(\text{NO}_x\) mixing ratios show a reduction of up to 0.5 ppbv (< 10\%) along the Indo-Gangetic Plain region and 0.2–0.3 ppbv (10–20\%) over the Thar Desert due to uptake of \(\text{NO}_2\) by dust particles. The changes in \(\text{NO}_2\) are also determined by the competition between the reactions of \(\text{NO}_2\) with dust and \(\text{OH}\) where the former tends to decrease \(\text{NO}_2\) while the latter tends to increase \(\text{NO}_2\) due to reduced \(\text{OH}\). The reduction in \(\text{NO}_x\) is in contrast with the results of Wang et al. (2012), who reported an increase in \(\text{NO}_x\) in the dust source region and attributed the increase to the renoxification process. Our study differs because Wang et al. (2012) assumed that renoxification process is active all the time, whereas this process is active only during daytime in our simulations. To quantify the contribution of renoxification process in the \(\text{NO}_x\) budget, we compared the \(\text{NO}_x\) distributions simulated by Dust_JH and Dust_JH_NoReNO4 configurations. The comparison (not shown) showed that the renoxification process does increase \(\text{NO}_x\) mixing ratios, but the magnitude of this increase (0.1–0.2 ppbv) is likely less than the reduction due to heterogeneous chemistry.

The spatial distributions of \(\text{HNO}_3\) and \(\text{H}_2\text{O}_2\) (Fig. 5) are largely modified by the dust storm with reduction of up to 2 ppbv (~99\%), because these species are highly reactive with dust particles (Table 1). The large reduction in \(\text{H}_2\text{O}_2\) mixing ratio estimated here is also in contrast with the results of Wang et al. (2012), who estimated an increase in \(\text{H}_2\text{O}_2\). Wang et al. (2012) assumed a 100\% yield for the conversion of \(\text{HO}_2\) into \(\text{H}_2\text{O}_2\) through heterogeneous uptake, while we assume a yield of 10\% following recent work by Bedjanian et al. (2013b). Mixing ratios of \(\text{NO}_x\), \(\text{N}_2\text{O}_3\) and \(\text{CH}_3\text{COOH}\) also show large reductions by Dust_JH with the highest decrease reaching up to 0.1 ppbv (~98\%), 0.46 ppbv (~99\%) and 0.45 ppbv (~96\%) respectively. The uptake of \(\text{HO}_2\) by dust particles leads to a maximum reduction of about 3.5 pptv (~40\%) over the Thar Desert region, which is much less than those reported previously (e.g., Bian and Zender, 2003; Wang et al., 2012) and is attributed to a lower γ value used in our study (~< 0.064 vs. 0.1). The maximum reduction in \(\text{OH}\), \(\text{CH}_2\text{O}\) and \(\text{CH}_3\text{OH}\) is estimated to be about 40, 21 and 5\% respectively.

The above discussion includes both the role of heterogeneous chemistry in changing the distribution of trace gases and dust-modified photolysis rate coefficients. The individual contributions of heterogeneous chemistry and perturbation in photolysis rate coefficients to the total difference in distributions of trace gases at the surface are estimated by comparing differences between simulations (Fig. 6). More
Figure 3. Spatial distributions of OMI-retrieved and WRF-Chem-simulated (Dust_JH and No_Dust) tropospheric column NO$_2$ during the low (13–16 April 2010) and high (17–22 April 2010) dust emission periods. WRF-Chem profiles are convolved with OMI averaging kernels before comparison. The percentage differences between low and high dust emission periods are also shown.

Figure 4. Spatial distributions of the WRF-Chem-simulated daytime (02:00–12:00 UTC) NO$_2$ photolysis rate with (Dust_J) and without dust (No_Dust) at the surface (top panel) and 100 hPa (bottom panel) during 17–22 April 2010 are shown. Percentage differences between Dust_J and No_Dust cases are also shown.
Figure 5. Spatial distributions of average surface O$_3$, SO$_2$, NO$_x$, HNO$_3$ and H$_2$O$_2$ mixing ratios simulated by WRF-Chem with Dust_JH and No_Dust configuration during 17–22 April 2010. The absolute difference between the two configurations is also shown. All values are in ppbv.

than 80% of the changes in distribution of these trace gases are explained by the heterogeneous chemistry. The changes in surface O$_3$ induced by dust-modified photolysis rate coefficients are within ±1–3 ppbv and are driven by the complex response of O$_3$ to decreases in photolysis rate coefficients. A decrease in the photolysis rate coefficients leads to slower photochemical processing in general and thus decreased O$_3$ production.

Mixing ratios of surface SO$_2$ and NO$_x$ show an increase of up to 0.1 ppbv due to reduction in photolysis rate coefficients. The increase in NO$_x$ and SO$_2$ is associated with a decrease in OH mixing ratios as reaction with OH is the main loss process for both NO$_x$ and SO$_2$. The decrease in O$_3$
Figure 6. Spatial distributions of absolute difference in average O$_3$, SO$_2$, NO$_x$, HNO$_3$ and H$_2$O$_2$ between Dust_JH and No_Dust (left panel), Dust_J and No_Dust representing the contribution of photolysis to total difference (middle panel), and Dust_JH and Dust_J representing the contribution of heterogeneous chemistry to total differences (right panel). All values are in ppbv and for the surface layer of the model.

Photolysis rate coefficient leads to a decrease of up to 30 % in OH (O$_3$ + h$_\nu$ → O$^1$D + O$_2$, producing OH through reaction of O$^1$D with water vapor) mixing ratios. Mixing ratios of surface HNO$_3$ and H$_2$O$_2$ also show a small decrease of up to 0.4 and 0.6 ppbv in the Dust_J configuration relative to the No_Dust configuration. HNO$_3$ is produced mainly by the reaction of OH with NO$_2$, but the rate of change of HNO$_3$ is dominated by changes in OH as the reduction in OH (up to 30 %) is larger than that in NO$_2$ (up to 5 %). The reduction in H$_2$O$_2$ is also associated with reduction in OH and HO$_2$ mixing ratios. Since OH is the major oxidizing agent in the troposphere, a decrease in OH also leads to a decrease in the oxidizing capacity of the atmosphere and a consequent increase of up to 5–10 % in several trace gases such as CO, alkanes and alkenes at the surface.
4.4 Impact of dust storm on vertical distribution of trace gases

In addition to changes in surface mixing ratios, vertical distributions of trace gases are modified by the dust storm via heterogeneous chemistry and dust-modified photolysis rates. By averaging over the region of maximum dust influence (70–80° E, 25–30° N) for each simulation, the percentage differences between Dust_JH and No_Dust giving the total difference, Dust_J and No_Dust giving the contribution of modified photolysis rates, and Dust_JH and Dust_J giving the contribution from heterogeneous chemistry are found. Profiles of O₃, NOₓ, HNO₃, OH and BIGALK are shown in Fig. 7. BIGALK represents alkanes with four or more carbon atoms in the MOZCART chemical mechanism and is shown to illustrate the dust-storm-induced changes in volatile organic compounds.

The changes in all the trace gases due to heterogeneous chemistry are much larger than those due to perturbations in photolysis rate coefficients, and are significant below 8 km, which is consistent with the vertical distribution of dust particles (Fig. 7). All gases except BIGALK show a net reduction because of the larger changes induced by the heterogeneous chemistry. The reduction in OH due to decrease in O₃ photolysis rate coefficient leads to a small increase of 1–4 % in NOₓ. The reaction with OH is the only loss process for BIGALK, and therefore BIGALK shows an increase due to both heterogeneous chemistry and perturbation in photolysis rates as both of these processes lead to a decrease in OH. The highest net decreases in O₃, NOₓ, HNO₃ and OH are estimated as ~16, ~26, ~91 and ~30 % respectively while the highest net increase in BIGALK is estimated as ~26 %. The vertical distributions of changes in NO₃, N₂O₅, H₂O₂ and CH₃COOH are similar to those in HNO₃, and the highest net reduction reaches 80–90 %.

4.5 Importance of RH dependence of reactive uptake coefficients (γ)

The uptake of trace gases by dust aerosols also depends upon the relative humidity as reactive uptake coefficients have a large variation with RH. The effect of relative humidity on the rate constants kO₃ and kHNO₃ at the surface over the model domain during 17–22 April 2010 is illustrated in Fig. 8, where model runs with and without RH dependence of γ are compared. The spatial distributions of average mass concentration for dust particles of 1.4 μm effective radius and relative humidity are also shown to help the interpretation. RH is less than 20 % over most of the Indian region and is 70–90 % over the oceanic regions. As expected, the spatial distributions of both kO₃ and kHNO₃ in both configurations are...
creases when RH effects on $\gamma$ and HO$_2$ dependence of $\gamma$ model layer between the model runs with and without RH differences in the distribution of these gases in the lowest for, resulting in a decrease in HNO$_3$ uptake coefficients increase when RH effects are accounted for these gases decreases with RH. In contrast, HNO$_3$ and H$_2$O$_2$ show a similar spatial distribution. The rate coefficients for other gases show a similar spatial distribution with RH while that with OH would decrease (increase) SO$_2$ if OH increases (decreases). The magnitudes of these RH-induced changes in trace gases are comparable to those induced by heterogeneous chemistry (reported in Sect. 4.3). This suggests that consideration of RH-dependent $\gamma$ values in heterogeneous chemistry calculations is as important as is the accurate simulation of dust mass concentrations.

4.6 Effect of uncertainty in reactive uptake coefficient ($\gamma$)

The effect of uncertainty in reactive uptake coefficient ($\gamma$) on heterogeneous chemistry induced changes in the tropospheric chemistry is illustrated in Fig. 10, where model runs with lower (Dust_JH_LoG) and upper (Dust_JH_HiG) bounds of $\gamma$ are compared in the lowest model layer. It is clear that uncertainty in $\gamma$ can lead to significant uncertainty in heterogeneous chemistry induced changes in all gases. Surface O$_3$ mixing ratios have uncertainties of up to 8–11 ppbv (25–30 %), while other gases have uncertainties of up to 1–1.5 ppbv (30–100 %), except SO$_2$ and NO$_2$, which show uncertainties of up to 3 ppbv (30–70 %) at some locations. A comparison of these uncertainties with the total change induced by dust storm in these gases (Fig. 5 and Sect. 4.3) reveals that uncertainties in O$_3$, HNO$_3$, NO$_3$, N$_2$O$_5$, OH, HO$_2$, H$_2$O$_2$, and CH$_3$COOH due to the uptake values have magnitudes similar to the total changes induced by dust aerosols, while those in NO$_2$, SO$_2$ and CH$_3$OH are even greater than the total changes induced by dust storm. These results highlight the importance and necessity of accurate measurements of reactive uptake coefficients.

4.7 Impact of dust aging

The aging of dust particles through heterogeneous uptake of gases can modify the dust reactivity towards trace gases. We have made an attempt to simulate changes in the dust reactivity and uptake of trace gases due to nitrate and sulfate coating, and the results are presented in this section. The spatial distributions of WRF-Chem-simulated average mass concentration of fresh and aged dust particles of 1.4 µm effective radius at the surface in Dust_JH_NO$_3$, Dust_JH_SO$_4$.
Figure 9. Spatial distributions of relative percentage differences in \( O_3 \), \( SO_2 \), \( HNO_3 \), \( H_2O_2 \), \( OH \) and \( HO_2 \) mixing ratios at the surface between Dust_JH and Dust_JH_NoRH configurations during 17–22 April 2010.

Figure 10. Spatial distributions of absolute differences in \( O_3 \), \( HNO_3 \), \( NO_2 \), \( SO_2 \), \( CH_3OH \) and \( CH_2O \) surface mixing ratios between Dust_JH.LoG and Dust_JH.HiG configurations during 17–22 April 2010. All values are in ppbv.
and Dust_JH_NO3_SO4 configurations during 17–22 April 2010 are shown in Fig. 11. In general, fresh dust particles are seen mostly in and near the source regions because all dust particles are emitted as fresh particles and have much smaller concentrations than the aged dust particles particularly outside the dust source regions. Among all the gases providing nitrate coating on dust particles, the HNO3 uptake makes the highest contribution to the concentration of aged particles due to its higher mixing ratios and strong increase in its reactivity towards the dust surface with relative humidity. For example, if there are 10 fresh dust particles cm⁻³ of effective radius 0.73 µm, then the uptake of HNO3, NO2, NO3 and N2O5 leads to about 2.5 aged dust particles cm⁻³ in one time step (180 s) at RH = 0, and HNO3 and NO3 each provide about 1 dust particle. However, the number concentration of aged dust particles formed due to HNO3 uptake increases to about 7 particles cm⁻³ per time step as RH increases to 20–40 %. The uptake of SO2 also increases with increase in relative humidity, but the SO2 contribution to the aged particles is much less than that of HNO3 due to lower SO2/γ values. The decrease in O3/γ values for nitrate-coated dust particles leads to an enhancement of 1–2 ppbv in surface O3 over the Thar Desert and western Indo-Gangetic Plain region relative to uncoated dust particles (Fig. 11) while increase in O3/γ values for sulfate-coated dust particles leads to a reduction of 0.5–1 ppbv over these regions. Since nitrate coating leads to a larger fraction of aged particles than the sulfate coating, surface O3 mixing ratios show an overall enhancement of up to 1 ppbv over the regions of high dust loadings when both sulfate- and nitrate-coated dust particles are allowed to react with O3 in the model.

The presence of a nitrate or sulfate monolayer on the dust particle might saturate the dust particles and deactivate them for further catalytic uptake of other gases. In order to examine the effect of such a coating on the uptake of O3, H2O2, HCHO and CH3COOH, lower and upper bounds of heterogeneous chemistry reaction rates induced changes in the surface mixing ratios of these gases are calculated (Fig. 12). Lower and upper bounds for each gas are calculated by subtracting their average values in Dust_J configuration from those in Dust_JH and Dust_JH_Sat configurations respectively. Absolute mixing ratios of these gases in Dust_J configuration, in which dust aerosols affected photolysis rates only, are also shown to provide an idea of the modification in base levels of these gases due to heterogeneous chemistry. As expected, the saturation of dust particles decreases the magnitude of reduction caused by heterogeneous chemistry for all the gases by 5–50 %. The amount of maximum reduction in surface O3 changed from 20–25 % to 15–20 % when saturation effects are accounted for. The saturation of dust particles has a larger impact on the distribution of H2O2 and CH3COOH as maximum reduction in both of these gases decreased to 20–40 % (as compared to 70–90 % for unsaturated dust particles) over the Thar Desert and dust source regions. Both of these gases show a small increase of 0.1–0.2 ppbv (less than 10 %) outside the source regions for the case of saturated dust particles. The maximum reduction in HO2 and OH (not shown) also changes from 20–40 % in Dust_J to 5–20 % in Dust_JH_Sat. Some changes can also be discerned in the distribution of HCHO and CH3OH (not shown), but they are within ±5 %.

5 Summary

The effects of a typical pre-monsoon season dust storm on tropospheric chemistry are analyzed for a case study in northern India, using the Weather Research and Forecasting model coupled with Chemistry (WRF-Chem), which is further developed to enhance its ability to simulate tropospheric chemistry in the presence of dust particles. The changes made to the model are specific to the MOZCART setup. Two major updates are included in this study: firstly, the F-TUV photolysis scheme of the model is updated to include the effect of dust aerosols on photolysis rates and to achieve consistency between the methods through which aerosols affect the meteorology and photolysis rates in the model. Secondly, a new scheme consisting of 12 heterogeneous reactions is included to simulate heterogeneous chemistry on the surface of dust particles. The relative humidity dependence of uptake coefficients, which was ignored in most previous studies, is used for six of the heterogeneous reactions.

The extended configuration of WRF-Chem is applied to a typical pre-monsoon season dust storm that occurred in northern India during 17–22 April 2010. The model reproduced the spatial and temporal distribution of dust plumes and aerosol optical properties (Kumar et al., 2014). The simulations are evaluated against surface O3 and NOy observations at a high-altitude (1958 m) measurement station in the Himalayan region (Nainital), and the model is found to capture the observed decrease in O3 and NOy during the dust storm only after the inclusion of the effects of dust on photolysis rates and heterogeneous chemistry. Average observed and modeled O3 values at Nainital during 17–22 April 2010 are estimated to be 56 ± 10 and 58 ± 5 ppbv respectively, and the corresponding NOy values are estimated to be 843 ± 887 and 1189 ± 751 pptv respectively. The extended configuration of the model also reduced biases in tropospheric column NO2 by up to 30 % compared to OMI retrievals and captured the general features of the dust-storm-induced changes in the spatial distribution of OMI-retrieved tropospheric column NO2.

Several sensitivity simulations are conducted to investigate the contribution of different processes on mixing ratios of several key trace gases including ozone, nitrogen oxides, hydrogen oxides, methanol, acetic acid and formaldehyde. The dust storm leads to a decrease of 5–25 % in photolysis rate coefficients of O3, NO2 and other trace gases at the surface and an increase of 1–5 % above 4 km. It is found to have a significant impact on the regional tropospheric chemistry: a
Figure 11. Spatial distributions of WRF-Chem-simulated average mass concentration of fresh and aged dust particles of 1.4 µm effective radius at the surface in Dust_JH_NO$_3$, Dust_JH_SO$_4$ and Dust_JH_NO$_3$.SO$_4$ configurations during 17–22 April 2010. $\Delta$O$_3$ is calculated by subtracting the average surface O$_3$ in Dust_JH configuration from the corresponding average values in Dust_JH_NO$_3$, Dust_JH.SO$_4$ and Dust_JH_NO$_3$.SO$_4$ configurations respectively.

decrease of 5–99% is estimated in the mixing ratios of a variety of trace gases including O$_3$, nitrogen oxides, hydrogen oxides, sulfur dioxide, methanol, acetic acid and formaldehyde at the surface. Analysis of the vertical distributions of these trace gases shows that dust-storm-induced changes are significant up to an altitude 8 km and are estimated as 80–90% (5–10 times) for highly reactive gases such as HNO$_3$, NO$_3$, N$_2$O$_5$, H$_2$O$_2$ and CH$_3$COOH. It is found that the majority of these changes are induced by the heterogeneous chemistry, while the contribution of dust-modified photolysis rates generally remained less than 10%. An increase of up to 30% in volatile organic compounds is estimated due to decreases in OH concentrations.

The RH dependence of $\gamma$ is found to play a large potential role in heterogeneous chemistry. Sensitivity studies showed that the exclusion of the RH dependence can introduce a difference of 1–2 orders of magnitude in heterogeneous reactions rate constants, 20–25% changes in O$_3$ and HO$_2$, and up to 50 and 100% changes in H$_2$O$_2$ and HNO$_3$, respectively. These effects are comparable to heterogeneous chemistry induced changes in these gases. The effect of uncertainties in the uptake coefficients ($\gamma$) on the distribution of trace gases is examined using sensitivity simulations with lower and upper bounds of $\gamma$. Differences of trace gas mixing ratios caused by uncertainties in the uptake coefficients are found to be the same magnitude or greater as the differences in mixing ratios induced by dust storm effects on tropospheric chemistry. We also tested the importance of atmospheric aging of dust particles in the context of heterogeneous chemistry. Model experiments based on laboratory studies of changes in dust reactivity due to atmospheric aging showed that coating of dust with nitrate and sulfate may lead to changes of up to 1 ppbv in surface O$_3$ simulations. A hypothetical simulation is also conducted by saturating the nitrate- and sulfate-coated
dust particles for uptake of O\(_3\), HO\(_2\), OH, H\(_2\)O\(_2\), CH\(_3\)COOH, CH\(_3\)OH and HCHO. The saturation of dust particles is found to have a larger impact on the distributions of H\(_2\)O\(_2\) and CH\(_3\)COOH but a relatively small impact on other gases.

This study clearly shows that the pre-monsoon season dust storm can potentially affect the regional tropospheric chemistry in northern India. However, the implications of the heterogeneous uptake of trace gases on aerosol size distributions and their feedbacks on the radiation budget and cloudiness are not examined here. Dust particles coated with nitrate/sulfate may interact differently with radiation as compared to uncoated dust particles and can increase or decrease cloudiness depending upon their size distribution. Both of these processes will have important implications for the direct aerosol radiative forcing and the elevated heat pump (EHP) hypothesis (Lau et al., 2006), which proposed that the absorption of solar radiation by dust and black carbon aerosols along the southern slopes of Himalayas modulates the meridional temperature gradient and leads to an early onset of Indian summer monsoon. The heterogeneous chemistry scheme implemented in the MOZART chemical mechanism here can be easily extended to a more detailed aerosol module (e.g., MOSAIC) of WRF-Chem, which would allow studies on the implications of heterogeneous chemistry for aerosols and their interaction with radiation, clouds, and the Asian monsoon, including the role of aerosol aging on aerosol and trace gas distributions.

Nevertheless, this study demonstrates that the effects of dust aerosols through heterogeneous chemistry and perturbation in photolysis rates should be included in atmospheric chemistry transport models, especially for simulating air quality in northern India. Although this study analyzed a typical dust storm in northern India, but more such studies should be conducted in future to lend further confidence in these results. At the same time, it is also imperative to improve the accuracy and precision of the reactive uptake coefficients, their dependence on relative humidity and atmospheric processing of dust particles. In addition, extensive efforts must be made to conduct co-located measurements of O\(_3\) and related gases, along with physical and chemical properties of dust aerosols in northern India, especially during the dust storm season, to gain insights into the effects of dust aerosols on tropospheric chemistry and provide more data for model evaluation.

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