



The anthropogenic contribution to atmospheric black carbon concentrations in southern Africa: a WRF-Chem modeling study

F. Kuik¹, A. Lauer^{1,a}, J. P. Beukes², P. G. Van Zyl², M. Josipovic², V. Vakkari³, L. Laakso^{2,3}, and G. T. Feig⁴

¹Institute for Advanced Sustainability Studies (IASS) Potsdam, Germany

²Unit for Environmental Sciences and Management, North-West University, Potchefstroom, South Africa

³Finnish Meteorological Institute, Helsinki, Finland

⁴South African Weather Service, Pretoria, South Africa

^anow at: Deutsches Zentrum für Luft- und Raumfahrt (DLR), Institut für Physik der Atmosphäre, Oberpfaffenhofen, Germany

Correspondence to: F. Kuik (friderike.kuik@iass-potsdam.de)

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Abstract. South Africa has one of the largest industrialized economies in Africa. Emissions of air pollutants are particularly high in the Johannesburg-Pretoria metropolitan area, the Mpumalanga Highveld and the Vaal Triangle, resulting in local air pollution. This study presents and evaluates a setup for conducting modeling experiments over southern Africa with the Weather Research and Forecasting model including chemistry and aerosols (WRF-Chem), and analyzes the contribution of anthropogenic emissions to the total black carbon (BC) concentrations from September to December 2010.

The modeled BC concentrations are compared with measurements obtained at the Welgegend station situated ca. 100 km southwest of Johannesburg. An evaluation of WRF-Chem with observational data from ground-based measurement stations, radiosondes, and satellites shows that the meteorology is modeled mostly reasonably well, but precipitation amounts are widely overestimated and the onset of the wet season is modeled approximately 1 month too early in 2010. Modeled daily mean BC concentrations show a temporal correlation of 0.66 with measurements, but the total BC concentration is underestimated in the model by up to 50 %.

Sensitivity studies with anthropogenic emissions of BC and co-emitted species turned off show that anthropogenic sources can contribute up to 100 % to BC concentrations in the industrialized and urban areas, and anthropogenic BC and co-emitted species together can contribute up to 60 % to PM₁ levels. Particularly the co-emitted species contribute signifi-

cantly to the aerosol optical depth (AOD). Furthermore, in areas of large-scale biomass-burning atmospheric heating rates are increased through absorption by BC up to an altitude of about 600 hPa.

1 Introduction

South Africa is one of Africa's largest economies and anthropogenic emissions of air pollutants from South Africa are of increasing concern. Due to South Africa's growing economy, fossil fuel consumption and energy demand are rising, with most of the electricity produced by coal-fired power plants (Lourens et al., 2011; Tiitta et al., 2014). Further contributions come from its large mining and metallurgical industry and domestic combustion, especially in informal settlements around towns (Venter et al., 2012). A large portion of South African anthropogenic emissions originate from the area around Johannesburg and Pretoria (Freiman and Piketh, 2003), a metropolitan area with a combined population of more than 10 million (Lourens et al., 2012), as well as the Mpumalanga Highveld and Vaal Triangle industrial areas that have both been declared pollution hotspots by the South African government (Lourens et al., 2011). Furthermore, large-scale biomass burning emissions contribute particularly during the dry winter season to air pollutant concentrations (e.g. Swap et al., 2003; Vakkari et al., 2014). The

issue of air pollution in South Africa has been recognized and explored in several recent studies, mainly focusing on the measurement and characterization of both gaseous species and particulate matter (e.g. Laakso et al., 2008, 2012; Vakkari et al., 2011, 2013; Venter et al., 2012; Jaars et al., 2014; Tiitta et al., 2014; Beukes et al., 2013).

So far, regional modeling studies for southern Africa have mainly focused on the meteorology (e.g. Cr  tat et al., 2011). Solmon et al. (2006) modeled aerosols, including BC, with a domain covering Europe and large parts of Africa but did not include South Africa. They identify poorly developed emission inventories – especially for Africa – as one of the main deficiencies in modeling aerosol concentrations and note a lack of measurement data for model evaluation in Africa. The African Multidisciplinary Monsoon Analysis (AMMA) project was designed to address these gaps. Part of it was dedicated to comparing the performance of chemical transport and chemistry climate models, simulating the distribution of trace gases and aerosols over West Africa. The findings provide recommendations for future air chemistry modeling, emphasizing the need for improved anthropogenic emission inventories (Ruti et al., 2011). Laakso et al. (2013) simulated particle growth in South Africa with an offline global aerosol model. They found that the model does not reproduce the observed particle formation characteristics. This was attributed mainly to the emissions, with their monthly resolution not capturing the emissions' variability. Tummon et al. (2010) simulated the direct and semi-direct effects of biomass-burning aerosols in southern Africa, reporting regional changes induced by aerosols including surface cooling and heating of the atmosphere in higher altitudes, leading to enhanced tropospheric stability and a decreased height of the planetary boundary layer. The study does not explicitly assess anthropogenically emitted aerosols. To the authors' knowledge, no peer-reviewed study exists to date that is aimed at modeling anthropogenic black carbon (BC) with a regional model over southern Africa.

Black carbon is an important component of air pollution. It is a carbonaceous aerosol that is produced during the incomplete combustion of carbon-based fuels and materials. It is an aggregate of rapidly coagulating small carbon spheres, with a total size generally below 1 μm . Black carbon is characterized by its strong absorption of visible and near-infrared light and by its resistance to chemical transformation (Ogren and Charlson, 1983; Goldberg, 1985; Petzold et al., 2013).

In addition to the burning of biomass and industrial processes, in particular domestic cooking and heating, as well as the transport sector are major sources of BC in Africa (Bond et al., 2013). Fine particulate matter, and thus BC contained within, is associated with several adverse effects on human health. These include respiratory and cardiovascular morbidity, such as aggravation of asthma, respiratory symptoms and an increase in hospital admissions, as well as mortality from cardiovascular and respiratory diseases and from lung cancer (Janssen et al., 2012). Some empirical studies suggest that

long-term exposure to $\text{PM}_{2.5}$ containing a high BC fraction may have larger mortality effects than other $\text{PM}_{2.5}$ mixtures (Smith et al., 2009).

The efficient absorption of solar radiation by BC makes these aerosols the most important absorber of visible light in the atmosphere. In addition to absorbing light while being suspended in the atmosphere, BC can reduce the amount of reflected sunlight when deposited on high albedo surfaces such as snow and ice. After carbon dioxide, emissions of BC are thought to make the second strongest contribution to current global warming (Ramanathan and Carmichael, 2008; Hodnebrog et al., 2014), though the exact climate forcing of BC is still under debate (e.g. Samset et al., 2014).

As BC has a short residence time in the atmosphere (few days) in comparison to CO_2 (several years up to more than 100 years), emission reduction measures would rapidly lead to a decrease in concentrations, which would have beneficial effects for both air quality and climate (Ramanathan and Carmichael, 2008; Shindell et al., 2012). These properties lead to BC often being classified as a Short-Lived Climate-forcing Pollutant (SLCP) (e.g. Schmale et al., 2014).

The metropolitan areas in South Africa are highly populated, and at the same time the population is highly vulnerable to air pollution and climate change because of their rather limited resources for adaptation. This is why an assessment of the contribution of anthropogenic BC emissions to the observed aerosol concentrations is needed as a first step for assessing potential emission-reduction scenarios. This is the aim of this study, which is done for the entire subcontinent of southern Africa, but with the main focus on the country of South Africa.

This study presents (Sect. 2) and evaluates (Sect. 3) a model setup for southern Africa, using the Weather Research and Forecasting (WRF) model online coupled to air chemistry and aerosol processes (WRF-Chem, Grell et al., 2005; Fast et al., 2006; Skamarock et al., 2008). The evaluation includes a comparison of the model results to different observational and reanalysis data. An important data set is the ground measurements conducted at Welgegund, ca. 100 km southwest of Johannesburg, detecting both pollution plumes coming from the industrialized and urban areas, as well as air masses representing the regional southern African background. It is one of the only regionally representative and comprehensive long-term inland atmospheric measurement stations (Beukes et al., 2013). In addition, data from observations of particulate matter ($\text{PM}_{2.5}$ and PM_{10}) and aerosol optical depth (AOD) are compared with the model results. With two sensitivity studies, the contribution of anthropogenic BC and co-emitted species to aerosol concentrations, AOD, and the impact on atmospheric heating rates is also analyzed (Sect. 4).

Table 1. General features of the setup, physics and chemistry schemes used in the configuration of the Weather Research and Forecasting model with chemistry (WRF-Chem).

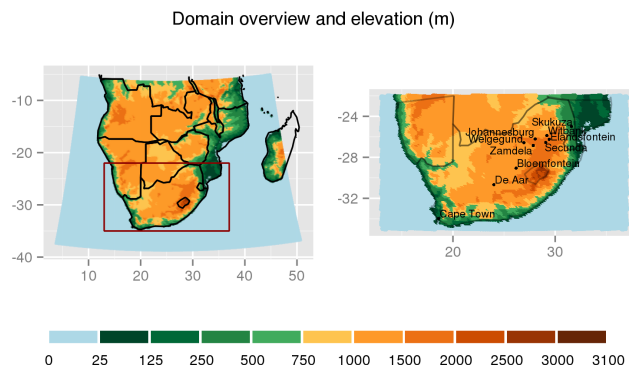
General features		
Domain size	4–50° E, 5–39° S	
Modeling time period	26 Aug–31 Dec 2010	
Resolution	15 × 15 km, 31 vertical levels (top at 10 hPa)	
Physics	Scheme	Remarks
cloud microphysics	Lin et al.	
radiation (shortwave)	Goddard	called every 15 min
radiation (longwave)	Rapid Radiative Transfer Model (RRTMG)	called every 15 min
boundary layer physics	Mellor–Yamada–Janjić (MYJ)	called every time step (90 s)
land surface processes	Noah LSM	
cumulus convection	Grell 3-D	called every time step (90 s)
chemistry	RADM2 with CMAQ aqueous chemistry	chem_opt = 41
photolysis	Fast-J	called every 15 min
aerosols	MADE/SORGAM	
dust	GOCART (online)	dust_opt = 3

2 Model and model simulations

2.1 Model description and setup

We apply the Weather Research and Forecasting model (WRF) version 3.5.1 (Skamarock et al., 2008) with chemistry and aerosols (WRF-Chem, Grell et al., 2005; Fast et al., 2006). We use the RADM2 chemistry scheme with the MADE/SORGAM aerosol module and aqueous phase chemistry (CMAQ) (Table 1). RADM2 in combination with the MADE aerosol module has already been widely used in literature (e.g. Grell et al., 2011; Minsenis and Zhang, 2010; Tuccella et al., 2012). Aqueous phase chemistry has been switched on as we expect this to be of relevance particularly when simulating aerosols during the wet season. The model has been set up with one domain covering large parts of southern Africa (4–50° E, 5–39° S, Fig. 1) at a horizontal resolution of 15 km × 15 km. WRF-Chem is configured with 31 vertical σ -levels, of which 14 levels are below 700 hPa. The model top is at 10 hPa.

We use the Modern-Era Retrospective Analysis for Research and Applications (MERRA) from the National Aeronautics and Space Administration (NASA) as initial and lateral atmospheric boundary conditions (Rienecker et al., 2011). The MERRA data with a horizontal resolution of $0.5^\circ \times 0.67^\circ$ at 6-h time intervals and at 32 pressure levels between 1000 and 10 hPa are interpolated to the model grid using the standard WRF Preprocessing System (WPS). European Centre for Medium-Range Weather Forecasts (ECMWF) Interim reanalysis (ERA-Interim) data are used as initial conditions for soil temperature and soil moisture (Dee et al., 2011). The modeled temperature, horizontal wind, humidity, surface pressure, and geopotential height are nudged to the lateral boundary conditions within a buffer zone of 5

**Figure 1.** Domain overview and elevation (left), enlarged for South Africa (right), locations of all stations with data used for the model evaluation (see text) and the location of Johannesburg included in the figure.

grid points normal to the lateral boundaries. This buffer zone is excluded in the analyses of the model results shown below. We prescribe sea surface temperatures (SSTs) using the National Oceanic and Atmospheric Administration (NOAA) optimum interpolation (OI) daily analysis (Reynolds et al., 2007). The SST data are based on daily mean satellite observations from the Advanced Very High Resolution Radiometer (AVHRR) and the Advanced Microwave Scanning Radiometer (AMSR) with a horizontal resolution of $0.25^\circ \times 0.25^\circ$. The diurnal SST variation is included in our SST forcing and is calculated following the surface energy budget method of Zeng and Beljaars (2005). Chemical boundary conditions for trace gases and particulate matter are created from simulations with the global chemistry transport Model for Ozone and Related chemical Tracers (MOZART-4/GEOS-5, Emmons et al., 2010).

The physics and chemistry modules used in the WRF-Chem configuration are summarized in Table 1.

2.2 Emissions

Anthropogenic emissions are taken from the EDGAR HTAP v2.2 inventory, released in fall 2013 (EDGAR: Emission Database for Global Atmospheric Research of the Joint Research Centre, JRC, of the European Commission, in cooperation with the Task Force on Hemispheric Transport of Air Pollution, TF HTAP, organized by the United Nations Economic Commission for Europe's Convention on Long-range Transboundary Air Pollution, LRTAP). The data set combines different available national or regional inventories. Gaps are filled by the bottom-up global emission inventory EDGARv4.3, which is calculated based on activity data and corresponding emission factors (see Janssens-Maenhout et al. (2012) for details on v1.0 of the data set; LRTAP-Wiki (2014) for updated information on v2). It should be noted that emission data for southern Africa are entirely based on EDGARv4.3, since currently no comprehensive regional inventories are available. EDGAR HTAP v2.2 reports monthly data for the emissions from the energy, industry, transport and residential sectors and annual data for emissions from shipping and aviation (only takeoff and landing included here). Emission data from small agricultural fires are not available in v2.2 and were therefore taken from v1.0 (annual data for 2005). The authors of the data set recommend using a satellite product for large-scale burning (Janssens-Maenhout et al., 2012), as indicated below.

For biomass burning emissions, the Fire Inventory from the National Center for Atmospheric Research (NCAR) version 1 (FINN, Wiedinmyer et al., 2011) is used. The data are based on daily satellite observations of fires and land cover, which are combined with emission factors and estimated fuel loadings. Fires of approximately 1 km² size are detected (Wiedinmyer et al., 2011). Biogenic emissions are calculated online by the Model of Emissions of Gases and Aerosols from Nature (MEGAN, Guenther et al., 2006).

2.3 Simulations

In this study, we performed a reference run (RR) with the model setup and emissions described above and two sensitivity runs (S1 and S2) with modified emissions but an otherwise identical setup. The model integration covers the time period from 26 August through 31 December 2010. Although September is the first spring month, it is still part of the dry season since rain usually occurs after mid-October. Black carbon concentrations in the interior of South Africa usually peak in September. In contrast, December 2010 is part of the wet season, which has significantly lower ambient BC levels. The first five days of all experiments were discarded as a spin-up period and excluded from the analysis presented.

2.3.1 Sensitivity studies

In the first sensitivity run (S1), all energy-related anthropogenic BC emissions and emissions from small agricultural fires are set to zero. Following Bond et al. (2013), energy-related emissions include all emissions from industry, transport (including aviation and shipping), energy production and residential heating. In addition, large scale biomass burning emissions are reduced to 35 % of the initial values. It is assumed that 65 % of the large-scale biomass-burning emissions are caused by humans directly or indirectly. This is based on estimates of the portion of pre-industrial biomass burning emissions of 37 % globally (Bond et al., 2013), as well as 33 % globally and 36 % for southern Africa (Dentener et al., 2006).

When aiming at reducing BC concentrations it is usually not feasible to only cut the BC emissions. If emissions from a certain source are reduced, usually also the emissions of co-emitted species such as sulfur dioxide and organic carbon are reduced. Those species can have a cooling impact on the climate. Hence, when assessing the maximum effect of cutting anthropogenic BC emissions on aerosol loadings, or the impact on meteorological variables, it is not sufficient to only consider a case without anthropogenic BC emissions. An integrated assessment of such emission cuts also needs to consider the contribution of the co-emitted components such as organic particles or sulfur dioxide (SO₂).

The above-mentioned second sensitivity simulation (S2) looks at the impact of both anthropogenic BC and co-emitted (often climate-cooling) aerosols. In addition to the reductions of BC (S1), also the emissions of co-emitted organic carbon (OC), primary sulfate aerosols (SO₄) and SO₂ are reduced in S2. The emissions are reduced in the same way as BC, i.e. the anthropogenic emissions are set to zero and the biomass burning emissions are reduced to 35 % of the original values. This is a simplifying assumption, as there might be sources among the anthropogenic source categories that do not emit BC but that do emit OC or SO₂ and vice versa. However, there is not enough information in the anthropogenic emission data used in this study (EDGAR HTAP v2.2) to make additional assumptions on sources emitting BC, but no OC or SO₂, or the other way round. For instance, the ratio of BC to OC emissions is constant throughout the whole model domain. Ideally, specific reduction factors should be employed. As there is no such information available for southern Africa, the above-described set up is used to estimate the overall effect of anthropogenic BC sources on aerosol loadings and atmospheric heating rates.

The anthropogenic contribution to aerosol concentrations, aerosol optical depth (AOD) and atmospheric heating rates is estimated as the differences between the reference run and the respective sensitivity simulations (S1, S2).

2.3.2 Model evaluation

For the model evaluation and a consistency check of the emissions, various observational and reanalysis data have been used (see Sect. 3).

A major data source for evaluating the model results from WRF-Chem is data obtained at the Welgegund measurement station (Fig. 1), the only long-term monitoring station measuring BC representative of the interior of South Africa (e.g. Venter et al., 2012; Vakkari et al., 2013; Tiitta et al., 2014). The station was set up at Welgegund in 2010 and is jointly operated by the North-West University (South Africa), the University of Helsinki and the Finnish Meteorological Institute. The station consists of an atmospheric monitoring trailer (Petäjä et al., 2013). Measured quantities used in this study include several meteorological parameters (temperature, relative humidity, wind speed and direction, and pressure), trace gases (SO_2 , NO/NO_x , O_3 , CO) and equivalent BC determined with a Multi-Angle Absorption Photometer (MAAP) corrected for the filter change artefact (Hyvärinen et al., 2013). There are no major anthropogenic pollution sources close to the station. Beukes et al. (2013) identified five important source regions for air masses analyzed at Welgegund. These include metallurgical industries in the Bushveld Igneous Complex (western and eastern limb) in the north to northeast of Welgegund, the Johannesburg-Pretoria megacity to the east, the Vaal Triangle (petrochemical, metallurgical and other industries) and the Mpumalanga Highveld (coal mining, coal-fired power plants, petrochemical operations and metallurgical smelters) between the east and south east. Furthermore, large-scale biomass-burning emissions originate mostly from the sector east of the measurement station (from north to south), because the biome in the western sector is drier and there is thus less plant material for combustion available from this sector. Additionally, the sector between north and south to the west of Welgegund is representative of the regional background of southern Africa.

In addition, observations of AOD from two AERONET-stations (Holben et al., 1998) and observations of $\text{PM}_{2.5}$ and PM_{10} provided by the South African Air Quality Information System (SAAQIS) of the South African Weather Service (SAWS) from three stations in the vicinity of the Pretoria-Johannesburg megacity are used (Fig. 1). The SAAQIS stations' main purpose is the monitoring of air quality in areas with high air pollution. The stations are classified as urban (Witbank station), residential (Zamdela station) and located in an urban residential area (Secunda station). As these are stations close to anthropogenic, non-biomass burning emission sources, aerosol concentrations are expected to be mainly dominated by local anthropogenic emissions. Hence, these comparisons are used as a consistency check for the emission data used here.

3 Comparison with observations and reanalysis data

3.1 Meteorology

3.1.1 WRF-Chem model results for southern Africa

The comparison of the sea-level pressure modeled with WRF-Chem to ERA-Interim reanalysis data (Dee et al., 2011) shows that WRF-Chem represents the common features of the southern African pressure distribution well (Fig. 2a). In September, over the south Atlantic and the south Indian Ocean, the edges of two subtropical highs can be identified. Another high is over the east coast of the continent, which is part of the high pressure belt around 30°S that influences the daily weather patterns of southern Africa (Tyson and Preston-Whyte, 2000). The spatial correlation between the WRF-Chem monthly mean and the ERA-Interim reanalysis monthly mean in September is high ($r = 0.95$) and the domain-averaged mean bias with respect to the ERA-Interim reanalysis is small (-0.6 hPa). In December, both WRF-Chem results and ERA-Interim reanalysis show that the low pressure area over the northern part of the model domain associated with the Intertropical Convergence Zone (ITCZ) is moving southwards compared to September, resembling the easterly low situation, which is usually the dominant synoptic situation in December (Tyson and Preston-Whyte, 2000). This constellation, associated with the ITCZ moving southwards at the beginning of the wet season, is responsible for strong precipitation over the subcontinent. The spatial correlation between the two data sets is slightly lower in December ($r = 0.79$) than in September. The domain-averaged monthly mean pressure of the WRF-Chem results is biased with respect to the ERA-Interim reanalysis by -2.6 hPa in December.

In September 2010, WRF-Chem simulated over the area within the northern low pressure region some precipitation, possibly indicating too early of an onset of the rainy season (Fig. 2b). Compared with the Global Precipitation Climatology Project (GPCP) precipitation data (Huffman et al., 2001), WRF-Chem overestimated the precipitation amounts in September 2010 as most parts of the subcontinent do not receive any significant amount of precipitation. The mean bias for the whole model domain is 1.11 mm day^{-1} . In December, the WRF-Chem results show large amounts of precipitation over the whole eastern part of the subcontinent, including Madagascar and the Mozambique Channel. WRF-Chem strongly overestimated the amounts of precipitation during all months of the modeled period, with a maximum monthly mean bias of $+6.47\text{ mm day}^{-1}$ (domain average) in December 2010 ($+200\%$). In addition to the northeastern part of the model domain and the Mozambique Channel, precipitation in the model was also strongly overestimated near the Drakensberg Mountains and over/on the edges of the South African Highveld. The spatial correlation is rather low in both September ($r = 0.38$) and December ($r = 0.36$).

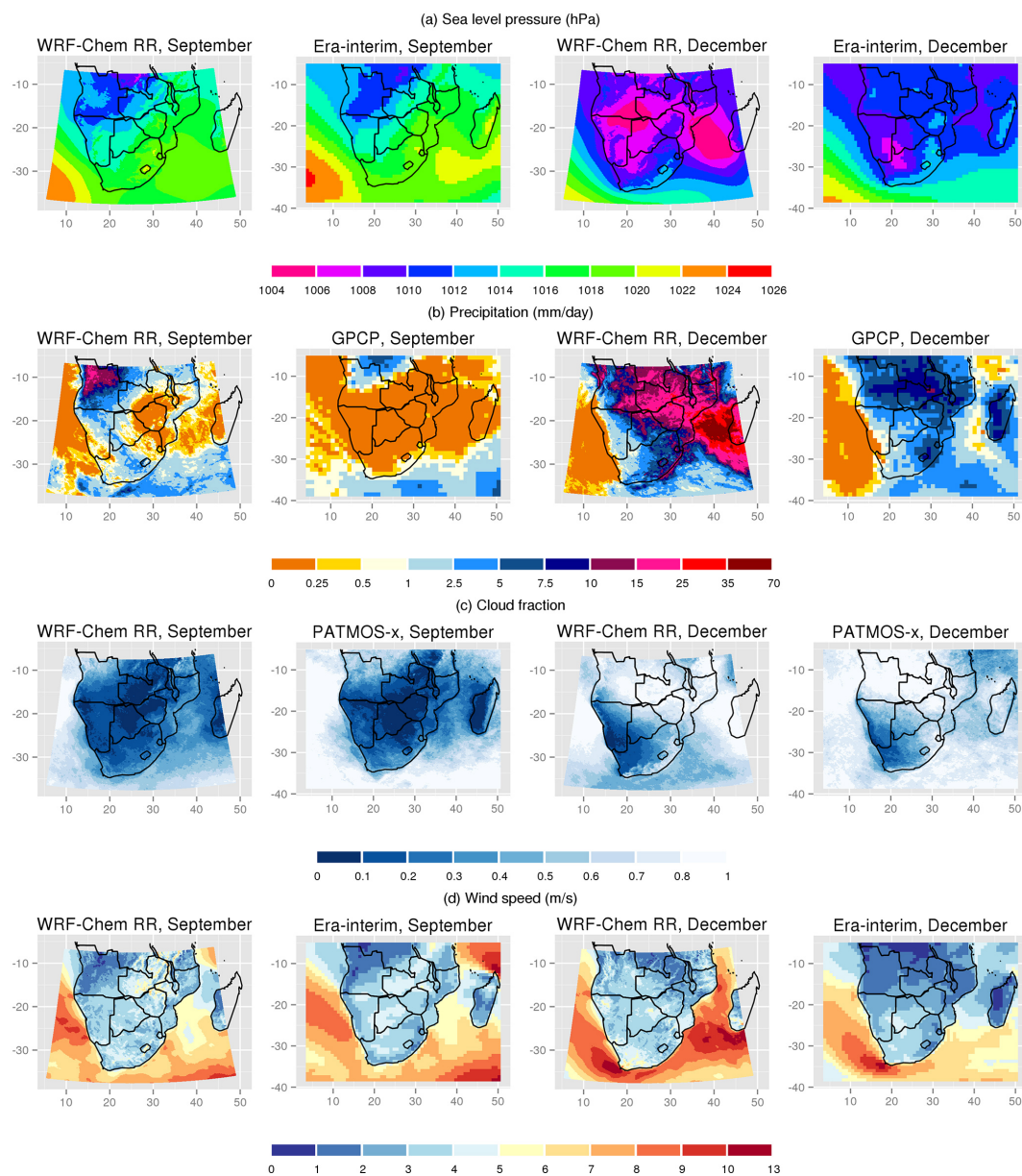


Figure 2. Selected meteorological variables, monthly means for September and December 2010, comparison of WRF-Chem model results with different data sets (**a** – sea level pressure, comparison with ERA-Interim reanalysis data, **b** – precipitation amount, comparison with GPCP data, **c** – cloud fraction, comparison with PATMOS-x satellite data, **d** – wind speed, comparison with ERA-Interim reanalysis data).

It should, however, be noted that satellite-based precipitation data sets such as the GPCP data over southern Africa also include uncertainties. It is known, for instance, that satellite-based precipitation estimates tend to underestimate rainfall amounts during the dry seasons (Huffman et al., 2001, 2007). Furthermore, underestimation of precipitation is documented for the GPCP data in areas of complex terrain (Huffman et al., 2001), which applies to the region around the Drakensberg Mountains and the edge of the Highveld, i.e. the escarpment.

Previous studies have shown that the extent and location of the model domain of a regional model is important when running simulations over South Africa (Cr  tat et al., 2011, and references therein). However, in a pre-study different configurations of WRF have been tested and the precipitation bias found in all simulations did not show any significant improvement when adjusting the extent of the model domain to, for instance, include Madagascar. Precipitation biases have been reported in the literature when studying South African meteorology with different models. Results suggest that this phenomenon might be related to too strong of an atmospheric

water cycle and too strong of an advection of moisture in the models. Another reason for the precipitation bias over the continent might be the complex topography (Crétat et al., 2011), as for example suggested by the overestimation near the Drakensberg mountains and the edge of the Highveld. Crétat et al. (2011) have shown that the precipitation bias in WRF depends partly on the choice of the cumulus scheme in combination with the PBL and microphysics schemes. However, their setup providing best results could not be used in combination with WRF-Chem as certain chemistry options such as aqueous phase chemistry are not available for all convection schemes.

In September 2010, most parts of the subcontinent including the northern parts of South Africa, Botswana and Zimbabwe were almost entirely cloud free (Fig. 2c). The modeled monthly mean cloud-free area is somewhat larger than the one obtained from the PATMOS-x satellite data (Heidinger et al., 2014). The cloud fraction in the model is particularly larger than observed in regions where precipitation is simulated, but the overall bias is negative with -13% (-27%) in September and -5% (-7%) in December. The spatial correlation between the two data sets is high ($r = 0.85$) in September. As already seen for sea level pressure and precipitation, the spatial correlation is smaller for December 2010 ($r = 0.57$), which might be related to more frequent and stronger convection in December, which is difficult to capture with a model.

The monthly mean 2 m temperature is modeled well in September 2010 (not shown), showing only a small domain averaged bias of 0.4°C compared with the ERA-Interim reanalysis data and a high spatial correlation ($r = 0.93$). In December, the modeled 2 m temperature is lower than the ERA-Interim reanalysis particularly in the areas above the continent where a positive precipitation bias is found. The overall mean bias of the model results compared to the ERA-Interim reanalysis in December is small with -0.03°C , and the spatial correlation is still high with ($r = 0.91$).

The outgoing longwave radiation (OLR, not shown) over some parts of the continent is up to 25 to 50 W m^{-2} lower than the PATMOS-x satellite data for September 2010. Again, the bias is particularly high in areas with a strong bias in precipitation, with a domain-averaged monthly mean bias of -21.4 W m^{-2} (-7%). The spatial correlation with the PATMOS-x data is $r = 0.85$. Likewise, in December, the modeled OLR over the Mozambique Channel is up to 100 W m^{-2} smaller than in the satellite data, with a domain-average bias of the monthly mean of -17.3 W m^{-2} (-7%) in December. The spatial correlation of the OLR is $r = 0.72$ in December 2010. The underestimation by the model in areas with large amounts of precipitation suggests that the cloud top heights are overestimated in WRF-Chem, even though the cloud fraction is underestimated. This in turn suggests that the cloud thickness is overestimated by the model, which could explain the stronger than observed precipitation. The cloud and precipitation biases can also explain the negative

bias in the surface temperature in regions with significant precipitation. The model biases of the different variables are consistent and might be the result of the difficulty in reproducing the observed convection, clouds and precipitation with WRF.

Compared with the ERA-Interim reanalysis, WRF-Chem captures the monthly mean 10 m wind speed (Fig. 2d) fairly well in the dry season (September), with some local positive, as well as negative deviations. The mean bias averaged over the whole model domain is -0.4 m s^{-1} (-8%) in September and 1.3 m s^{-1} (30%) in December. The spatial correlation is fairly high with $r = 0.87$ in September and $r = 0.85$ in December.

3.1.2 Comparison to measurements at Welgegund

For comparing the model results to measurements done at the Welgegund station, the modeled daily means of all variables considered except precipitation are averaged over the 3×3 nearest grid points surrounding the measurement station. The WRF-Chem precipitation data are compared to TRMM satellite data (Huffman et al., 2007) and averaged over the 25 (5×5) nearest grid points to be comparable to the TRMM nine grid point average. All the comparative data are presented in Fig. 3 and Table 2.

The TRMM data show that precipitation events become more frequent from mid-October 2010 on, with almost no precipitation observed beforehand. From this, we qualitatively derive the beginning of the rainy season around mid-October 2010. In contrast, the beginning of the rainy season in the model is about 1 month too early. In addition, the amplitudes of the precipitation events are much higher, at times up to three times as high as the TRMM values (e.g. in mid-December with ca. 20 mm day^{-1} as indicated by the TRMM data and more than 60 mm day^{-1} in WRF-Chem). The modeled time series of the precipitation in September is not correlated with the TRMM data ($r = -0.09$). In December, there is a rather low correlation ($r = 0.20$), and a mean bias of 2.17 mm day^{-1} (44%).

The monthly mean specific humidity measured at Welgegund increases from 5.39 g kg^{-1} in September to 11.86 g kg^{-1} in December 2010. This increase is captured qualitatively by the model, but the WRF-Chem results are positively biased in September ($+0.96\text{ g kg}^{-1}$, $+18\%$), October and November, with the bias decreasing over time. In December 2010, the modeled monthly mean specific humidity is in good agreement with the observations with a mean bias of -1.5% (-0.18 g kg^{-1}). The temporal correlation between modeled and measured daily mean values is high in September ($r = 0.76$). While the monthly mean is modeled well in December 2010, the time series of the daily values are uncorrelated ($r = 0.06$), meaning that the model does not capture the day-to-day pattern.

The monthly mean 2 m temperature at Welgegund does not vary much over the whole modeling period, with 19.4°C

Table 2. Monthly mean of modeled and observed meteorology, BC and selected trace gases, mean bias of the WRF-Chem daily means with respect to observations and temporal correlation coefficient (Pearson) of daily means over 1 month or the whole period September to December 2010 (all). Observational data are station measurements at Welgegend for all variables except for precipitation, which is obtained from the TRMM satellite data.

Variable	Month	WRF-Chem mean and standard deviation	Measurements mean and standard deviation	Mean bias	Correlation coefficient
Welgegend					
Specific humidity (g kg^{-1})	September	6.35 ± 2.09	5.39 ± 2.00	0.96	0.76
	December	11.76 ± 1.67	11.86 ± 1.78	−0.18	0.06
	All	9.67 ± 2.98	8.80 ± 3.29	0.87	0.78
Precipitation (mm day^{-1})	September	1.0 ± 2.4	0.001 ± 0.002	1.0	0.09
	December	7.1 ± 12.9	4.9 ± 5.2	2.2	0.20
	All	5.6 ± 10.7	2.1 ± 4.3	3.5	0.26
2 m temperature ($^{\circ}\text{C}$)	September	19.6 ± 2.4	19.4 ± 2.4	0.2	0.84
	December	20.0 ± 1.6	21.0 ± 2.0	−1.0	0.44
	All	19.8 ± 2.1	20.6 ± 2.6	−0.9	0.69
10 m wind speed (m s^{-1})	September	4.8 ± 1.7	5.0 ± 1.6	−0.2	0.77
	December	3.9 ± 1.4	6.1 ± 1.2	−2.2	−0.07
	All	4.3 ± 1.6	5.6 ± 1.6	−1.3	0.29
BC concentration ($\mu\text{g m}^{-3}$)	September	0.73 ± 0.35	1.47 ± 0.70	−0.74	0.62
	October	0.43 ± 0.29	0.88 ± 0.50	−0.45	0.67
	November	0.32 ± 0.20	0.31 ± 0.13	0.01	0.11
	December	0.25 ± 0.15	0.19 ± 0.11	0.06	0.37
	All	0.43 ± 0.31	0.71 ± 0.67	−0.28	0.66
CO (ppb)	September	170 ± 41	201 ± 60	−31	0.78
	December	100 ± 15	115 ± 17	−14	0.40
	All	131 ± 38	161 ± 60	−29	0.78
O ₃ (ppb)	September	42 ± 7	50 ± 10	−8	0.82
	December	29 ± 6	34 ± 4	−5	0.14
	All	34 ± 8	42 ± 10	−8	0.73
SO ₂ (ppb)	September	1.8 ± 2.1	1.8 ± 1.2	−0.04	0.31
	December	2.6 ± 2.3	1.0 ± 0.9	1.6	0.44
	All	2.2 ± 2.4	1.3 ± 1.1	0.9	0.21
NO _x (ppb)	September	4.6 ± 5.4	3.0 ± 1.8	1.7	0.75
	December	9.2 ± 8.4	2.5 ± 1.2	6.7	0.36
	All	6.8 ± 7.4	3.4 ± 1.6	3.4	0.19

measured in September and 21.0°C in December. It is simulated quite well in September, with only a slight mean warm bias of $+0.2^{\circ}\text{C}$. From October on, the modeled monthly means are biased slightly negative. In December, the mean bias is -1.0°C . As is evident from Fig. 3, as well as a comparison of the standard deviations (SDs, not shown), the model captures the day-to-day variability well. From October on, the timing of the modeled minima and maxima, as well as the minimum and maximum values agree less well with the observations. This is also seen in the temporal correlation coefficient of the daily means, which decreases from September ($r = 0.84$) to December ($r = 0.44$). The temporal correlation over the whole period is $r = 0.69$.

The monthly mean 10 m wind speed measured at Welgegend varies between 5.0 m s^{-1} (September) and 6.1 m s^{-1} (December). The model results are biased negatively in all 4 months, with the smallest bias in September (-0.2 m s^{-1} , -4%) and the largest bias in December (-2.2 m s^{-1} , -36%). From the beginning of October, the model has some difficulty in capturing the maximum daily mean wind speeds well, and underestimates the minimum daily mean wind speeds. The temporal correlation coefficient indicates that the daily mean wind speed is captured better in September ($r = 0.77$), while the model results and the measurements are uncorrelated in December ($r = -0.07$).

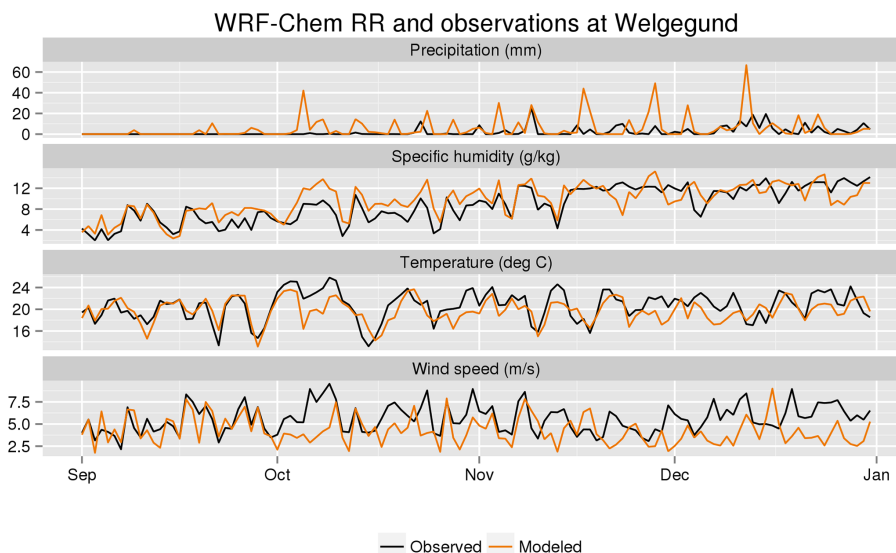


Figure 3. WRF-Chem model results of meteorological variables at Welgegund in comparison with Welgegund station measurements and precipitation satellite data (TRMM). Shown are daily means (daily sums in the case of precipitation).

Averaged over the whole modeling period, the modeled 10 m wind direction deviates slightly from the measured wind direction at Welgegund (not shown). While the frequency of northerly winds – the dominant wind direction – is modeled quite well (around 27 %), the portion of wind coming from the northeast is clearly underestimated (less than 10 % in the model compared with around 20 % in the measurements). In contrast, the northwesterly wind direction is slightly overestimated. Wind coming from the southwest to the southeast does not play a major role at Welgegund, which is correctly reproduced by the model. The overestimation of wind coming from the northeast by the model is particularly prevalent in September (not shown). In November the modeled main wind direction is shifted only slightly to the east compared with the observations. The wind direction bias in October and December is small.

3.1.3 Atmospheric profiles and inversion layer height

For an analysis of the simulation of the atmospheric vertical structure, WRF-Chem temperature and humidity profiles were compared with available radiosonde measurements at Pretoria, Bloemfontein, de Aar and Cape Town (MetOffice, 2006, see Fig. 1 for the location of the stations). As data availability for 2010 is sparse, average profiles of measurements obtained between the years 1997 and 2012 are used for comparison. This is done as a consistency check to see whether the model is able to reproduce the main climatological features of the vertical profiles. The radiosonde measurements are not directly comparable to the WRF-Chem model results. The comparison shows that WRF-Chem is able to capture the basic (climatological) features of the vertical profiles of temperature and humidity with the modeled vertical

profiles being within the variability given by two times the standard deviation (2σ -range, not shown).

In addition to the average temperature and humidity profiles, the inversion layer height has been calculated from each measured vertical profile and from WRF-Chem at the times corresponding to the radiosonde ascents. The inversion height is determined according to the following criteria, following Cao et al. (2007):

- An inversion is characterized by increasing temperature with height and decreasing relative humidity.
- If present, the inversion is located between 825 and 350 hPa (inland stations) and between 950 and 600 hPa in Cape Town (elevation 42 m). The lowest couple of hundred meters are excluded to exclude radiative inversions at the surface.
- Only inversions with temperatures above 0 °C are searched for in order to avoid artifacts caused by falling ice particles.
- If there are several inversions within one profile, the inversion with the largest decrease in relative humidity is chosen.

In Pretoria the monthly mean inversion height varies between 740 and 710 hPa between September and December (Fig. 4). The inter-annual variability given by the 25th and the 75th percentiles lies between 800 and 650 hPa. The mean inversion height is slightly lower in September and October than in November and December. Similar behavior is also found at the other three radiosonde stations. The mean inversion height modeled with WRF-Chem at the four stations is generally slightly lower by about 50 hPa than the mean inversion

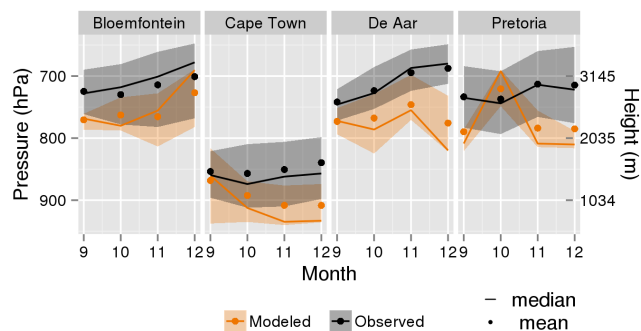


Figure 4. Inversion heights for each month (September–December) and four different radiosonde stations, lines represent the median, shaded areas the 25th and 75th percentiles, dots the mean values.

height obtained from the radiosonde measurements (mostly between ca. 800 and 750 hPa for the inland stations, with few exceptions). A comparison of the frequencies of measured and modeled inversions suggests that WRF-Chem might underestimate the number of days with an inversion present. We discuss the role of the inversion layer height for near-surface concentrations of BC in Sect. 3.2.3.

3.2 Black Carbon

3.2.1 Modeled monthly mean concentrations

Figure 5 shows the modeled monthly mean near-surface BC concentrations for September, October, November and December 2010, with “near-surface” meaning the lowest model layer, centered around about 30 m above the ground. The highest monthly mean BC concentrations in September are modeled in the Johannesburg-Pretoria area with values up to $15 \mu\text{g m}^{-3}$. In Zimbabwe, where the emission inventory also shows relatively high anthropogenic emissions, the BC concentrations are comparable to the Johannesburg-Pretoria levels (up to $2.5 \mu\text{g m}^{-3}$). In the north of the model domain at the border between Zambia and Angola mean BC concentrations are as high as $5 \mu\text{g m}^{-3}$ with biomass burning being the main BC source in that area. Over land, the lowest modeled monthly mean BC concentrations are found in the southeast of South Africa in the dryer Karoo regions, with values of less than $0.1 \mu\text{g m}^{-3}$. This region is relatively far from both anthropogenic sources and from large-scale biomass burning areas.

It can also be seen from Fig. 5 that the mean modeled concentrations are generally much higher in September 2010, which corresponds to the end of the dry season in the model, than in the following months. Especially in November and December, concentrations are lower, possibly due to a combination of higher removal of BC from the atmosphere (wet scavenging), the lack of large scale biomass burning as a major source and a less stable atmosphere (i.e. a smaller number of days with an inversion).

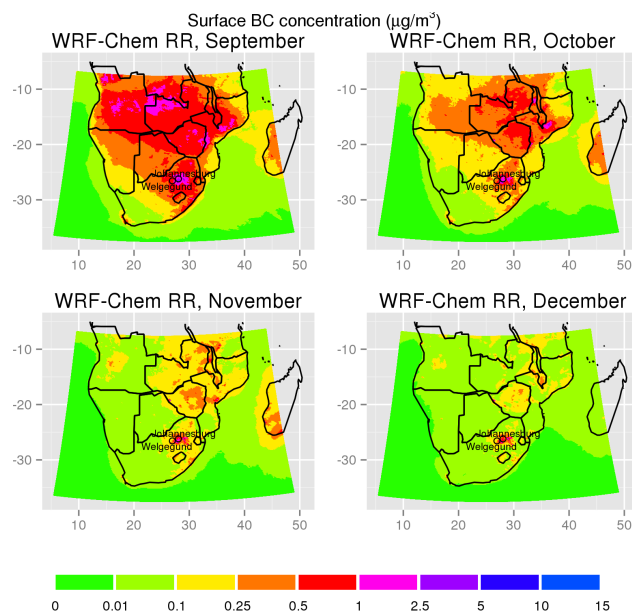


Figure 5. Monthly mean near-surface BC concentrations (lowest model layer) modeled with WRF-Chem, September–December 2010.

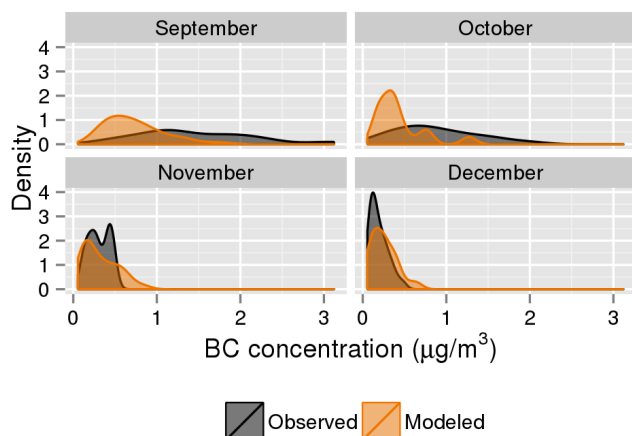


Figure 6. BC concentrations at Welgegend, measured and modeled with WRF-Chem: probability density functions (PDFs) for September–December 2010. The PDFs are calculated from the observed 15-min values and the 3-hourly values (instantaneous values) from the model results.

For comparison, the measured annual mean in Berlin, Germany ranges between around $2 \mu\text{g m}^{-3}$ at urban background stations and around $3.5 \mu\text{g m}^{-3}$ at measurement sites close to busy roads (2012 values, Senatsverwaltung für Stadtentwicklung und Umwelt, 2013). BC concentrations are especially high in some regions in Asia, e.g. in Kathmandu, Nepal, with an annual mean measured as $8.4 \mu\text{g m}^{-3}$ (Sharma et al., 2012).

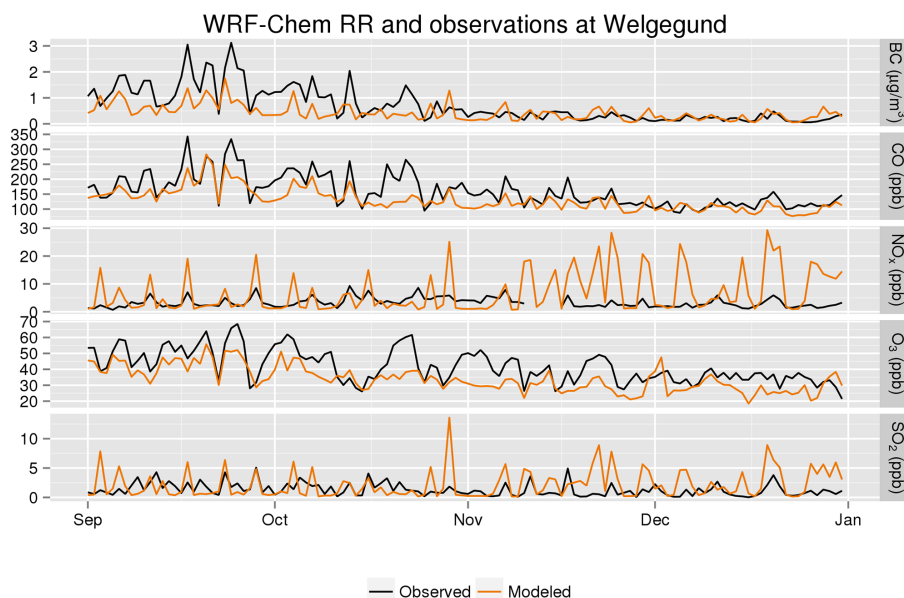


Figure 7. Comparison of modeled daily means of BC and gaseous species with station measurements at Welgegund.

3.2.2 Comparison with Welgegund data

At Welgegund the measured monthly mean BC concentrations decrease steadily from September (dry season) to December (wet season), with $1.47 \mu\text{g m}^{-3}$ in September, $0.88 \mu\text{g m}^{-3}$ in October, $0.31 \mu\text{g m}^{-3}$ in November and $0.19 \mu\text{g m}^{-3}$ in December (Table 2). The maximum daily mean concentrations in September are about $3 \mu\text{g m}^{-3}$ and in October about $2 \mu\text{g m}^{-3}$. As is evident from Fig. 6 the observed probability density function (PDF) for the mostly dry month of October is similar to the one for September, while the PDFs for the wetter months of November and December are much narrower and have distinct peaks at BC concentrations below $0.5 \mu\text{g m}^{-3}$.

The monthly BC means modeled with WRF-Chem are generally smaller than those shown by the measurements in September and October ($0.73 \mu\text{g m}^{-3}$ and $0.43 \mu\text{g m}^{-3}$, corresponding to a bias of -50 and -51%) and slightly higher than observed in November ($0.32 \mu\text{g m}^{-3}$, biased by 3%) and in December (just above $0.25 \mu\text{g m}^{-3}$, biased by 32%). Over the whole period, the mean bias is negative ($-0.28 \mu\text{g m}^{-3}$, -39%). The modeled PDFs in September and October are too narrow and the peaks around ca. $0.5 \mu\text{g m}^{-3}$ are at concentrations too low compared with the measurements. The modeled PDF for October resembles rather a wet season PDF than a dry season PDF, which is in line with the results we described for the simulated precipitation, showing that the beginning of the wet season is modeled ca. 1 month too early.

Even though the magnitude of the peak values and the average of the daily mean time series are underestimated in September and October, the time series of modeled and measured daily means are reasonably well correlated (tempo-

rally) with correlation coefficients of $r = 0.62$ (September) and $r = 0.67$ (October) (Fig. 7 and Table 2).

As the Welgegund station is not directly surrounded by sources of BC, apart from smaller local grass fires, most of the BC measured at Welgegund is transported to the station (Tiitta et al., 2014). Thus, the BC concentrations at Welgegund are strongly impacted by how effectively pollutants are transported from the industrialized areas, as well as from the biomass-burning areas mainly located in the sector east of the station. 96 h back-trajectories of air masses at Welgegund (Beukes et al., 2013) show that anthropogenic BC can be transported to Welgegund in different ways: either directly, with wind at Welgegund coming from the northeast to east, or by air masses re-circulated over the continent with wind at Welgegund from the north.

The pollution roses shown in Fig. 8 give a first estimate for the direction from which BC is transported to the station. The measurements show that very high concentrations are most frequently observed during periods with wind from the north or northeast corresponding to the above-mentioned transportation pathways. These pathways are reproduced by the model, which simulates the highest BC concentrations with wind coming from the northern to eastern sectors. However, as previously mentioned and visible from Fig. 8, the main wind directions in the model are shifted from the northeast to the northwest.

3.2.3 Discussion

Several factors are likely to influence the modeled BC concentration, including the bias in modeled meteorology (e.g. precipitation, wind direction), a low quality of the emis-

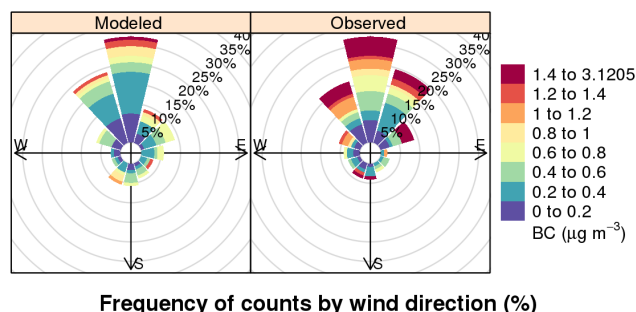


Figure 8. Pollution rose at Welgegund, comparison of WRF-Chem model results and station measurements. The plot shows the BC concentration modeled/measured for wind coming from the indicated directions and is created from the non-averaged data, e.g. 15-min values for the observations and 3-h values for the model results.

sion inventories, the choice of chemical boundary conditions or uncertainties and limitations in the representations of important processes in the model (e.g. the particle size distribution, the parametrization of convection or the boundary layer).

A too-early beginning of the rainy season and an overestimation of the precipitation amounts are likely to result in a too-strong wet deposition of aerosols including BC in the model and are likely two reasons contributing to the underestimation of the modeled mean BC concentrations particularly during the dry season at Welgegund. This is especially the case in October 2010, being mostly dry in the observations, but showing significant precipitation in the model. As BC has a typical atmospheric residence time of a few days, a full quantitative analysis on the impact of the overestimation in precipitation on modeled BC concentrations would require back-trajectories for several days, which is beyond the scope of this study. We argue qualitatively that the modeled overestimation of precipitation might contribute to the modeled underestimation of BC, as we know that BC has to be transported to the measurement site, because there are no significant sources close by.

Highest BC concentrations are modeled with wind from the northern to eastern wind sectors, which is consistent with the measurements at Welgegund. However, the shift in the modeled main wind direction to the northwest compared with the measurements likely also contributes to the above-discussed model bias in the BC concentrations. This is especially the case in September when the northeastern component of the wind is underestimated in the model, which is the second most frequently observed wind direction at Welgegund in this month. BC peak concentrations are measured particularly during these wind episodes.

Furthermore, the negative bias in modeled wind speed at Welgegund might also contribute to an underestimation of BC transported to Welgegund. However, this bias is fairly small and is likely not a main reason for the underestimation of modeled BC during the dry season.

The lack of BC transported from the industrialized and urban areas to Welgegund in September being a reason for the underestimation of modeled BC at the measurement station is further supported by the plots in Fig. 5 showing the geographical distribution of the modeled BC: higher BC concentrations resulting from urban emissions are found downwind of Pretoria and Johannesburg, while the Welgegund station is located just outside the area of the urban pollution plume with typical concentrations between 1 and $2.5 \mu\text{g m}^{-3}$ inside the plume. When comparing the BC concentrations measured at Welgegund to the model results at an equivalent location of Welgegund situated downwind of the modeled main wind direction at the same distance from the urban areas around Johannesburg and Pretoria as the Welgegund site (not shown), model and measurements are in much better agreement during the entire simulation period: in September, the modeled mean BC concentration at the equivalent location is above $1 \mu\text{g m}^{-3}$, and around $0.5 \mu\text{g m}^{-3}$ in October, reducing the model bias to values between -30 and -40% . This further supports that the modeled meteorology plays an important role in explaining the model bias of the BC concentration at Welgegund.

In principle, the height and strength of inversion layers can also influence the BC concentrations. A too-low number of inversion days in the model, i.e. an underestimation of days with stable atmospheric conditions, could result in pollutants being too-well mixed and in concentrations being too small. A too low inversion height in the model would increase the concentrations in the boundary layer during the inversion events and might counteract some of this. While the inversion height is captured quite well in the model, the number of inversion days is probably underestimated. The scarcity of radiosonde data in the fall of 2010 does not allow for a more detailed analysis of the inversion height statistics and a comparison of the modeled BC concentrations on days with and without inversion layers during the dry season. However, inversion layers are not thought to play a dominant role for the BC concentrations measured at Welgegund as the concentrations are dominated by transport processes over at least 100 km to the station allowing for ample mixing. This is supported by the finding that BC concentrations at Welgegund do not show a distinct diurnal cycle in September 2010 (not shown), which would be expected if the inversion played a significant role.

In general, the modeled meteorology and the modeled BC time series agree reasonably well with the observations at Welgegund during the dry season. A major contribution to the lower correlation of modeled meteorology, as well as the BC daily means with the Welgegund measurements during the wet season is likely caused by the difficulty of the model in reproducing the observed convection activity, which plays a major role particularly during the wet season.

Emission inventories of energy-related emissions of BC for Africa have rather large uncertainties (Bond et al., 2013). This certainly plays an important role for the modeled BC

concentrations, as the modeled concentrations can only be as good as the emission data used as model input. For example, day-to-day variations in BC concentrations due to the variability of energy-related anthropogenic BC emissions cannot be represented by the model as the emission inventory used (EDGAR HTAP) has a time resolution of one month or less. However, the analysis in this study suggests that the energy-related anthropogenic emissions are at least within the correct order of magnitude, as further elaborated in Sect. 3.3. Furthermore, the FINN biomass burning emission inventory with a time resolution of one day seems to capture the biomass burning events relatively well. This is suggested by the fairly good temporal correlation of the modeled daily means of BC with the measurements during the dry season, as biomass burning episodes play an important role for high levels of BC at Welgegund during the dry season (Tiitta et al., 2014).

Black carbon at Welgegund is measured as equivalent BC. It is therefore possible that BC is overestimated as additional non-BC absorbing material is also classified as BC. Studies disagree on the exact amplitude of the measurement uncertainty of the MAAP ranging from very little increase in absorption due to non-absorbing coatings of BC particles (e.g. Lack et al., 2012; Cappa et al., 2012, 2013) to a factor of two (e.g. Shiraiwa et al., 2010; Wang et al., 2014). However, we do not believe that the measurement uncertainty alone could explain a bias of 50 % in the dry season, but rather a combination of the model deficiencies and uncertainties discussed above.

3.3 Aerosol optical depth and particulate matter concentrations

3.3.1 Aerosol optical depth

Compared to the MODerate Resolution Imaging Spectroradiometer (MODIS) (Remer et al., 2005) satellite observations (MODIS Terra and Aqua monthly level-3 data, collection 5.1) of the aerosol optical depth (AOD), WRF-Chem captures the main geographical pattern over southern Africa qualitatively correctly, as exemplarily shown for September (Fig. 9), with high AOD values (larger than 0.3) in the northwest of the model domain, where biomass burning is strong, and a lower AOD in South Africa (mostly between 0.1 and 0.3).

Especially in the northwest of the model domain over the ocean the model results deviate strongly from the MODIS data (up to 90 %). The biases could be caused by several reasons which make a quantitative comparison difficult. In order to conduct a thorough quantitative evaluation of the model results with the satellite data, the model would have to include sampling of the data as seen from the satellite (e.g. taking into account the cloud cover and the specific satellite overpass times). This could not be done here. Furthermore, the uncertainty of the satellite data that can be quite large

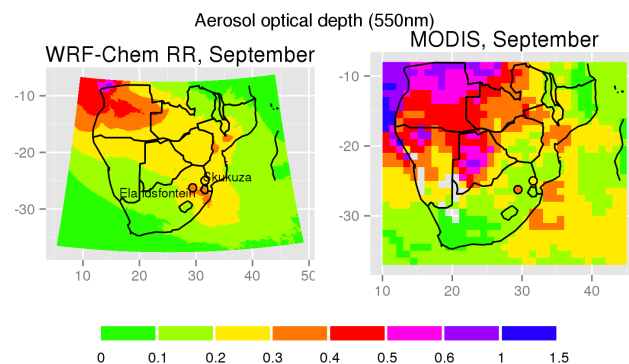


Figure 9. Comparison of modeled AOD with MODIS satellite observations, September 2010.

particularly for large AOD values (Ruiz-Arias et al., 2013) would have to be taken into account. This can also be seen in Fig. 9 showing ground-based AOD measurements from the AERONET network for comparison.

We therefore also compare modeled monthly mean aerosol optical depths with AERONET measurements at Skukuza, located in the Kruger National Park on the eastern border of South Africa, and Elandsfontein, located in the industrialized Highveld east of Johannesburg (Holben et al., 1998, see Fig. 1 for the location of the stations). Here, only measurements obtained under cloud-free conditions are used. Daily mean AODs are not available for every day from September 2010 to December 2010, with 10 missing days at Elandsfontein in September, 14 missing days at Skukuza in each October and November, and 19 missing days at Skukuza in December. The measured AODs at 500 and 675 nm are linearly interpolated to the AOD at 550 nm, which is calculated by the model.

The mean AOD (Table 3) is higher at Elandsfontein, which is located closer to anthropogenic aerosol sources, than at Skukuza. AOD is modeled reasonably well at both stations and during most months. Measured (modeled) means at Elandsfontein amount to ca. 0.31 (0.32) in September, 0.40 (0.54) in October, 0.21 (0.35) in November and 0.15 (0.40) in December, and at Skukuza to ca. 0.30 (0.25) in September, 0.32 (0.34) in October, 0.20 (0.21) in November and 0.14 (0.14) in December. Overall, the comparisons of the model results with the AERONET AOD show a reasonably good performance of WRF-Chem in simulating the AOD at this location. The fairly good agreement of the model results with measurements close to anthropogenic sources (Elandsfontein) suggests that total energy-related anthropogenic aerosol emissions are at least within the correct order of magnitude.

3.3.2 Particulate matter

The model results are further compared with the measurements conducted at stations of the South African Weather

Table 3. Monthly mean and mean bias of modeled and observed AOD, PM_{2.5} and PM₁₀.

Variable	Month	WRF-Chem mean	Measurements mean	Mean bias
Elandsfontein				
AOD	September	0.32	0.31	0.01
	October	0.54	0.40	0.14
	November	0.35	0.21	0.14
	December	0.40	0.15	0.25
Skukuza				
AOD	September	0.25	0.30	−0.05
	October	0.34	0.32	0.02
	November	0.21	0.20	0.01
	December	0.14	0.14	−0.0001
Secunda				
PM ₁₀ ($\mu\text{g m}^{-3}$)	September	49.17	143.56	−94.39
	October	51.45	68.72	−17.27
	November	38.10	38.52	−0.42
	December	32.41	38.55	−6.41
PM _{2.5} ($\mu\text{g m}^{-3}$)	September	45.03	46.49	−1.46
	October	48.30	25.94	22.36
	November	35.53	16.42	19.11
	December	30.16	17.52	12.64
Witbank				
PM ₁₀ ($\mu\text{g m}^{-3}$)	September	51.96	76.88	−24.92
	October	47.18	39.95	7.23
	November	34.47	22.30	12.17
	December	43.35	28.82	14.53
PM _{2.5} ($\mu\text{g m}^{-3}$)	September	46.33	36.34	9.99
	October	42.97	22.19	20.18
	November	31.00	13.37	17.63
	December	39.54	17.28	22.26
Zamdela				
PM _{2.5} ($\mu\text{g m}^{-3}$)	September	48.69	36.94	11.75
	October	51.56	34.89	16.67
	November	36.02	28.65	7.37
	December	35.01	29.00	6.01

Service (see Sect. 2) from September 2010 to December 2010, including Secunda (PM₁₀ and PM_{2.5}), Witbank (PM₁₀ and PM_{2.5}) and Zamdela (PM_{2.5}); as presented in Table 3 (see Fig. 1 for the location of the stations).

Averaged over the whole modeling period and all stations, WRF-Chem underestimated PM₁₀ by −26 % (observed: 58.42 $\mu\text{g m}^{-3}$, modeled: 43.50 $\mu\text{g m}^{-3}$), and overestimated PM_{2.5} by 51 % (observed: 27.02 $\mu\text{g m}^{-3}$, modeled: 40.84 $\mu\text{g m}^{-3}$). This could indicate that the size-distribution of primary particles such as mineral dust assumed in the model for the emissions of these particles might be too small.

WRF-Chem underestimated the PM₁₀ concentrations in September at Witbank up to −66 %. It is biased positively in

October (6 %), November (31 %) and December (43 %). At Secunda, a slight negative bias is found during all 4 months, from −32 % in September to only −1 % in November. The PM_{2.5} concentrations are – given the large uncertainties and model deficiencies as discussed for BC in Sect. 3.2.3, such as the low quality of emission inventories – modeled reasonably well for September at all three stations, with the modeled values biased for Witbank +28 % and Zamdela +32 % and only biased by −3 % in Secunda. For October, November and December the modeled concentrations at Secunda and Witbank are positively biased, with both the modeled range of daily means and the median being higher than the measurements.

The bias is smaller at Zamdela, especially during the wet season.

The biases might suggest that the different sources of PM might not be represented correctly in the emission data, or that the assumed particle size distributions are not representative for southern African conditions.

3.4 Gaseous species at Welgegund

To further assess the performance of WRF-Chem, results are compared with measurements at Welgegund (Fig. 7) for ozone (O_3), sulfur dioxide (SO_2), nitrogen oxides ($NO + NO_2 = NO_x$) and carbon monoxide (CO). The statistics including bias and temporal correlation coefficients are summarized in Table 2.

WRF-Chem has a negative bias in CO (-31 ppb/ -15% in September, -14 ppb/ -13% in December, -18% overall) and ozone (-8 ppb/ -15% in September, -5 ppb/ -15% in December, -19% overall). The modeled SO_2 is in good agreement with the measurements (bias -0.04 ppb, -2%) in September, but overestimated in December (1.6 ppb, $+160\%$). Averaged over the whole period the modeled bias in SO_2 at Welgegund is 0.9 ppb (65%). Likewise, the modeled NO_x is overestimated throughout the entire simulation period with biases ranging from 1.7 ppb (56%) in September to 6.7 ppb (270%) in December. The model bias of CO and O_3 is rather similar throughout all modeled months, while that of NO_x and SO_2 is much higher in December (wet season) than in September (dry season).

Particularly very high modeled NO_x values in December are probably related to very high emissions, which are higher than the maximum values found in the EDGAR HTAP emission inventory over Europe. This supports that emission inventories for Africa still have large uncertainties particularly for individual species and source regions.

In addition, we have compared model results for NO_2 (tropospheric column) and CO (lowest model layer) with satellite data (not shown). These qualitative comparisons show that the emission hotspots seem to be in the right locations.

3.5 Summary and conclusions from the model evaluation

The evaluation of WRF-Chem with ground observations, satellite data and the comparison to reanalysis and model data has highlighted some points that need improvement but also showed that overall both meteorology, aerosols and gaseous species are simulated reasonably well during the dry season, given the large uncertainties in, for instance, the emission data or the lateral boundary conditions as observations are generally very sparse in this region. Concerning the meteorology, a bias in precipitation exists with precipitation amounts being overestimated by the model particularly during the wet season over the Indian Ocean between Madagascar and continental East Africa as well as in the ITCZ. The

comparison of the model with measurement data obtained at the Welgegund measurement site confirms that precipitation amounts are mostly overestimated and that the beginning of the rainy season in the model is about 1 month too early (mid-September instead of mid-October). Furthermore, the main modeled wind direction at Welgegund is shifted towards the north which directly affects the modeled transport of atmospheric pollutants from the Johannesburg-Pretoria area towards the measurement station.

As for the modeled BC concentration at Welgegund, it is biased low in comparison to the measurement data in the dry season. The main reasons for this underestimation are likely the shift in main wind direction in the model, as well as the modeled early beginning of the rainy season, likely leading to enhanced wet deposition. Both of these shortcomings are expected to result in less BC transported to the measurement station than shown by the observations. This shows the importance of capturing the observed meteorology with the model in addition to reasonable emission estimates. An evaluation of a large-scale model with only a few available comprehensive measurement stations is challenging and underlines the need for further comprehensive monitoring sites in southern Africa. Especially the lack of comprehensive measurement stations in the western part of South Africa makes the model evaluation challenging. The effort of setting up further monitoring sites is underway (see Sect. 5).

The reasonably good temporal correlation of the BC daily means time series with measurements suggests that the biomass burning emissions, with a 1-day resolution, capture the biomass burning events reasonably well. The comparison of measured AOD, PM_{10} , $PM_{2.5}$ with model results in near-source regions further suggests that the total energy-related anthropogenic aerosol emissions in these regions seem to be within the correct order of magnitude. This might, however, not necessarily be true for individual species such as NO_x .

Overall, the qualitative reasonably good results as well as the identification of plausible reasons for the low bias of modeled BC at Welgegund suggest that the model setup is suitable for a first assessment of the contribution of anthropogenic BC and co-emitted species to aerosol concentrations on a regional scale and their impact on meteorology.

In addition to the above-discussed uncertainties in the model, model parameterizations and model parameters such as assumed particle size-distributions might not be well suited for application in this region. We therefore consider the results of this study on the anthropogenic contribution to BC concentrations in southern Africa as a very first and rough estimate and as a potential basis for comparison with future studies using improved models and better input data.

4 Contribution of anthropogenic BC sources to aerosol loadings

4.1 Black Carbon

4.1.1 Near-surface concentrations

The sensitivity run (S1) shows that anthropogenic emissions are the main contributors to BC loadings in many parts of the southern Africa (Fig. 10a). In September, anthropogenic BC contributes between 90 and 100 % to the simulated BC loadings in the center of the industrialized and urban area around Johannesburg and Pretoria, and the contribution is similarly high in coastal areas (especially around Cape Town), which are generally more populated than the south-western inland areas. In these coastal regions, savannah fires do not contribute significantly to the BC concentrations compared with the northern part of the subcontinent. At Welgegund, the share of anthropogenic BC in September ranges between 80 and 90 % (i.e. up to $5 \mu\text{g m}^{-3}$ in September and up to $2.5 \mu\text{g m}^{-3}$ in December). Energy-related anthropogenic emissions do not seem to play a large role in areas with strong biomass burning, where the share of BC concentrations caused by total anthropogenic emissions (60–70 %) is in the same range as the assumed fraction of anthropogenic biomass burning emissions (65 %). In December the anthropogenic portion of BC is up to 100 % in an area covering most parts of eastern South Africa and Zimbabwe, including Welgegund. In coastal areas the results for December are similar to the September result.

4.1.2 Vertical distribution

The mean BC differences in September are analyzed further at two latitudinal cross sections displaying the vertical profile of BC (Fig. 11): a northern cross section averaged over the latitudes 14.25 to 12.75°S , and a southern cross section averaged over 27.25 to 25.75°S . In order to reduce the noise, the data have, in addition to the monthly averaging, also been binned into 45-km bins consisting of three grid cells in the longitudinal direction. The vertical coordinate (pressure) is divided into 12 bins, each 50 hPa, the averaging time is 1 month. The northern cross section covers areas with strong biomass-burning emissions, and the southern cross section includes the Johannesburg-Pretoria megacity.

The cross sections show distinct differences between the two source regions of BC (Fig. 11a): in the northern domain, high BC concentrations of up to $0.5 \mu\text{g m}^{-3}$ are found up to ca. 500 hPa. It can be seen further that on average the plume does not rise much higher than 500 hPa, but is then rather transported out onto the South Atlantic Ocean. This is consistent with a persistent stably stratified 500-hPa layer described by Tyson and Preston-Whyte (2000). The cross section over the industrialized Highveld shows a different picture: the anthropogenic BC contribution decreases more rapidly with

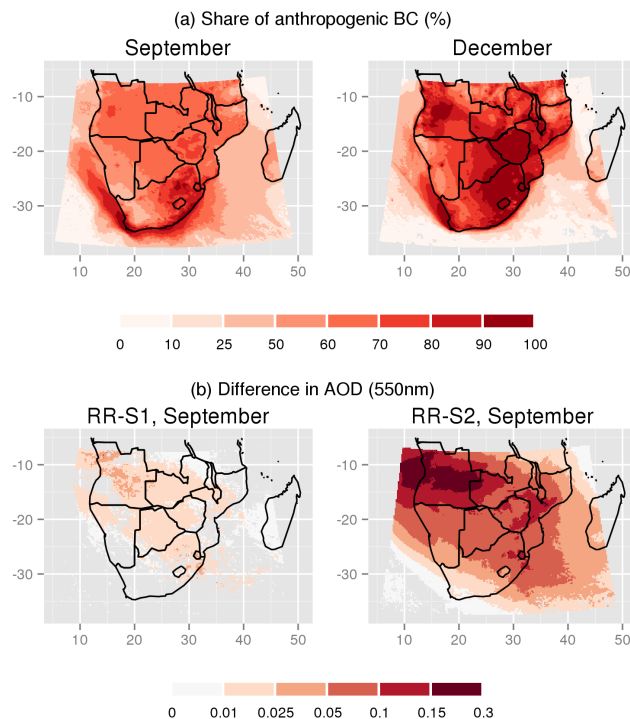


Figure 10. (a) Contribution of anthropogenic BC sources to BC concentrations, (b) contribution of anthropogenic BC sources to AOD (left: contribution of anthropogenic BC only, right: contribution of anthropogenic BC and co-emitted aerosols). For (b), the model results have been interpolated to a lon-lat-grid of $0.2^\circ \times 0.2^\circ$, and only grid cells statistically significant at a confidence level of 95 % are shown.

height, and the highest concentrations are found near the surface (the elevation of Pretoria is about 1300 m a.s.l.).

The anthropogenic contribution to BC concentrations ranges up to 90 % in the urban core of the southern cross section (Fig. 11b). In particular in this highly industrialized area the anthropogenic contribution to BC loadings is large and dominates the modeled concentrations. This is important when assessing for instance the health impact of (anthropogenic) BC. The share of anthropogenic BC ranges between 60 and 70 % in the biomass-burning area, both at the surface and at higher layers.

4.1.3 Biomass burning vs. energy-related emissions

The total share of BC from biomass-burning emissions (natural and anthropogenic) is estimated from the sensitivity run S1 by scaling up the modeled BC concentrations by a factor of $100\%/35\% = 2.86$. The scaled concentrations are then compared to the reference run (RR). In the urban and industrialized areas (averaged over $1.5^\circ \times 1.5^\circ$ around the metropolitan area of Johannesburg and Pretoria), the estimated contribution of biomass-burning emissions to the total near-surface BC concentrations amounts up to 62 %

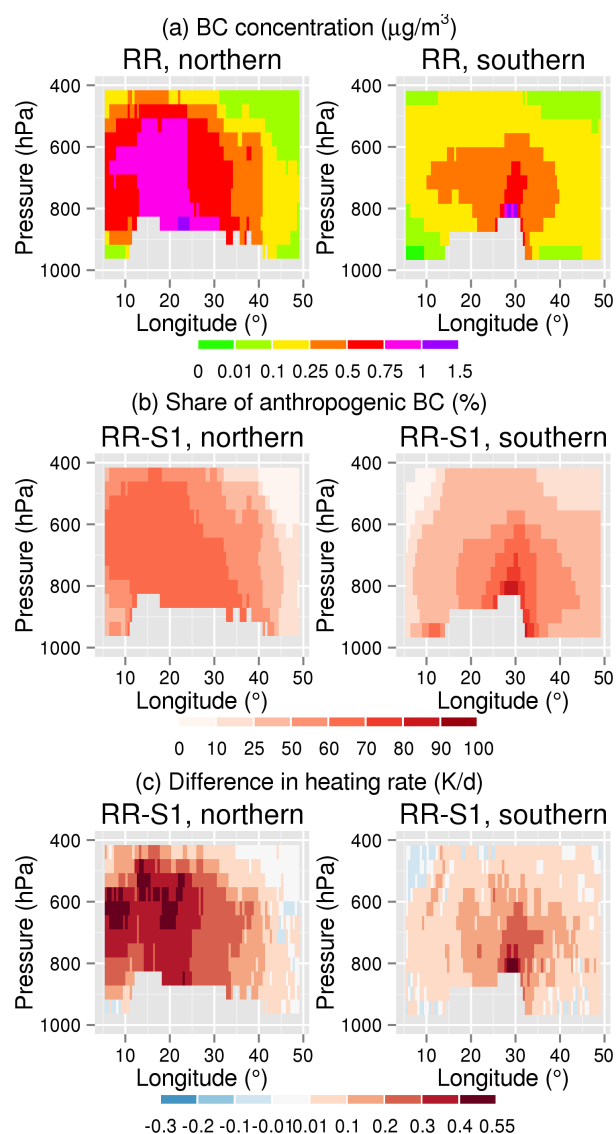


Figure 11. Vertical BC distribution (a), anthropogenic contribution to BC concentrations (b) and contribution of anthropogenic BC to atmospheric heating rates (c). All figures show the monthly mean results for September 2010.

(Fig. 12), but with much lower average values (25 % in September, 16 % in October, 5 % in November, 4 % in December). The model results further suggest that the contribution of biomass burning to the total BC is much higher at Welgegund, with monthly averages of 57 % in September, 44 % in October, 16 % in November and 10 % in December, confirming the findings of Tiitta et al. (2014) that biomass burning plays an important role for the BC levels observed at Welgegund during the dry season.

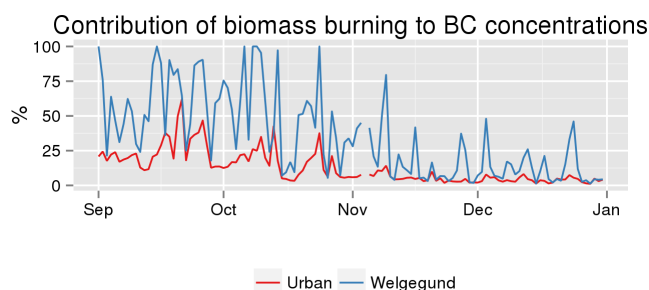


Figure 12. Estimated contribution of biomass burning emissions to BC concentrations in the Johannesburg-Pretoria urban area and at Welgegund.

4.2 Particulate matter and aerosol optical depth

BC particles are usually in the sub-micron size range (e.g., Petzold et al., 2005; Schwarz et al., 2008; Kondo et al., 2011) contributing only little to $\text{PM}_{2.5}$ and PM_{10} as these are often dominated by other particle types. In the following, we therefore focus on the contribution of BC to PM_1 .

In September, the PM_1 concentration modeled with WRF-Chem (not shown) reaches peak values of up to $55 \mu\text{g}/\text{m}^3$ around Johannesburg and Pretoria, and up to $30 \mu\text{g}/\text{m}^3$ in areas of highest biomass-burning emissions. In the northern part of the continent and in the surroundings of Johannesburg and Pretoria, modeled concentrations range mostly in between 10 and $15 \mu\text{g}/\text{m}^3$. In December the modeled PM_1 concentrations are highest around Johannesburg and Pretoria with values up to $30 \mu\text{g}/\text{m}^3$.

In the northern areas dominated by biomass burning the contribution of anthropogenic BC to the modeled (near-surface) PM_1 concentration in September ranges between 5 and 7.5 % with some spatial variations. The contribution of the modeled anthropogenic BC to PM_1 ranges up to 10 to 15 % in the surroundings of the Johannesburg-Pretoria area and between 7.5 and 10 % at Welgegund. Averaged over the whole urban area around Johannesburg and Pretoria, the mean contribution of anthropogenic BC to PM_1 in September amounts to 6 %.

The measured contribution of (total) BC to PM_1 at Welgegund is 13 % (average over one year, Tiitta et al., 2014). This value is not directly comparable to the model result for September but within a similar range as that calculated by the model for anthropogenic BC (10 %). Despite the underestimation of the absolute BC concentrations by the model both model results and measurements suggest that anthropogenic BC is an important contributor to PM_1 . This is especially important when assessing the health effect of PM.

When also accounting for the co-emitted species (OC and SO_2), the modeled contribution of both BC and co-emitted species to PM_1 in September is highest around Johannesburg and Pretoria amounting up to 60 %, underlining the impor-

tance of co-emitted species when assessing the contribution of anthropogenic BC sources to PM_{10} concentrations.

The difference in AOD between the reference run and the sensitivity simulation S1 (Fig. 10b) shows a similar spatial variability in September as PM_{10} . Only those grid cells with the differences significantly different from 0 at a 95 % confidence level are shown. The differences in AOD range between 5 and 10 % in both areas of strong biomass burning and over the industrialized Highveld. In December, the differences in AOD are not statistically significant and longer model integrations would be needed to improve the signal-to-noise ratio.

When also accounting for co-emitted species (S2) the contribution to AOD ranges up to 0.3 (50 %) in the northwest of the model domain. The contribution decreases towards the southeast, with an island of high contribution to AOD near Pretoria and Johannesburg (up to 0.15). In December, the total anthropogenic contribution of BC and co-emitted species to AOD is highest over Johannesburg and Pretoria with a maximum of up to 0.3.

The increase in AOD results in a reduction of incoming solar radiation at the surface or surface dimming (not shown), but the integration time was too short for the results to be statistically significant. This result is, however, consistent with what has been reported in the literature (e.g. Tummon et al., 2010).

4.3 Contribution of anthropogenic BC to atmospheric heating rates

The vertical heating rate difference (ΔHR , only the short-wave contribution is considered here) is assessed for September along the same cross sections discussed above (Fig. 11c). Atmospheric heating by anthropogenic BC in the model reaches up to ca. 0.7 K day^{-1} around 600 hPa in the northern cross section and near the surface around Johannesburg and Pretoria in the southern cross section. The strongest change in heating rates are spread vertically up to 600 hPa in the northern cross section, because emissions from biomass burning are efficiently transported to these heights in the model. In contrast, the maximum changes in heating rates caused by anthropogenic BC in the southern cross section are found close to the surface over the industrialized Highveld.

These model results suggest that OC and SO_2 co-emitted with anthropogenic BC play only a minor role for the heating rates compared with BC itself. The peak values in heating rates caused by anthropogenic BC and co-emissions (S2) amount to about 0.7 K day^{-1} , which are in the same range as modeled for anthropogenic BC alone (S1).

5 Conclusions

This study presents and evaluates a model setup for studying air chemistry and aerosol processes and their impacts

in southern Africa. In addition, a consistency check on the emission input data is done by comparing PM measurements with the model results in urban regions that are expected to be dominated by anthropogenic emissions. It then assesses the contribution of anthropogenic BC and co-emitted species to aerosol concentrations (BC and PM_{10}) and to aerosol optical depth, and assesses the impact of BC sources on atmospheric heating rates.

The evaluation of the WRF-Chem model applied over southern Africa shows that the main features of the meteorology such as temperature and sea-level pressure are modeled reasonably well, but some parameters, such as precipitation, are more problematic. Precipitation is very challenging to model: for example, Cr  tat et al. (2011) show that WRF has difficulties in reproducing observed precipitation amounts and patterns over southern Africa for a variety of different physics options.

Black carbon monthly mean concentrations are underestimated at Welgegund in September and October by ca. 50 %. Reasons contributing to the underestimation during the dry season are likely the biases in the modeled meteorological variables resulting in less than observed BC transported from the industrial source regions to Welgegund. These are in particular the shifted wind direction in September, the early beginning of the rainy season in the model in October and the overestimation of precipitation amounts. The importance of the wind direction bias for the BC bias is supported by the fact that the amplitude of the modeled monthly mean BC concentration is closer to the measured means when comparing to the results for an equivalent location (same distance from the urban and industrialized areas of Johannesburg and Pretoria as Welgegund but downwind of the modeled main wind direction). In November and in December the monthly mean BC concentrations at Welgegund are small and in reasonably good agreement with the measurements.

Besides the modeled meteorology, the high uncertainties in the emission inventories, the choice of chemical boundary conditions or uncertainties and limitations in the representations of important processes in the model (e.g. the particle size distribution, the parametrization of convection or the boundary layer) are likely to contribute significantly to the model biases in BC concentrations.

The modeled BC concentrations at Welgegund have a temporal correlation with measurements of 0.62 and 0.67 in September and October, respectively. This reasonable correlation can be attributed to the well-modeled day-to-day variability of the meteorology. This also suggests that the temporal resolution and pattern of the biomass burning emissions, which contribute significantly to the total BC at Welgegund, are a reasonable estimate of the real biomass burning emissions.

The comparison of the model results for AOD, $\text{PM}_{2.5}$, and PM_{10} with AERONET data and observations in the industrialized Highveld and Vaal triangle region, as well as the model qualitatively capturing the geographical pattern of the AOD

retrieved from MODIS satellite data, suggests that the magnitudes of the energy-related anthropogenic aerosol emissions used here (EDGAR HTAP) are, despite the generally low quality of emissions inventories for South Africa, a reasonable first estimate of the emissions. This is, however, not necessarily true for individual species such as NO_x or other source regions.

Two large source regions for anthropogenic BC are the industrialized and urban areas on the South African Highveld around Johannesburg and Pretoria (including the Mpumalanga industrial Highveld and the Vaal Triangle), and areas of large-scale biomass burning in the dry season. These are also the areas where the modeled BC concentrations are the highest on the subcontinent. These concentrations are strongly influenced by anthropogenic BC sources contributing up to 100 % in the industrialized and urban areas around Johannesburg and Pretoria both during the dry and the wet season. Biomass burning BC contributes only little to the total BC modeled in the area around Pretoria and Johannesburg but contributions increase significantly towards the outskirts of the area, e.g. at Welgegund.

An analysis of the atmospheric heating rates shows a slight surface heating over the industrialized South African Highveld (S1) that might be largely canceled by cooling due to surface dimming caused mainly by co-emitted species (S2). Black carbon from biomass burning at higher layers of the atmosphere leads to increased atmospheric heating rates and local warming in the lower and middle troposphere, possibly impacting the atmospheric stability in this region.

The high computational cost of WRF-Chem at 15 km resolution over the entire southern African subcontinent did not allow for long-term simulations. Instead, it was chosen to simulate only 4 months from September 2010 to December 2010, which include parts of the dry season, the wet season as well as the transition period. Because of the strong convection during the wet season, the signal-to-noise ratio of the model results is low particularly in December. Integrating over longer time intervals, spanning several seasons or years, would help increase the signal-to-noise ratio, thus increasing the robustness of the obtained results. Possible refinements of the model include a more detailed specification of the particle size distributions used for the emissions, ideally based on long-term measurements of the aerosol size-distribution in different source regions (natural and anthropogenic aerosols). Furthermore, future studies could assess whether a nudging to meteorological observational/reanalysis data would improve the model results, or the usage of urban parametrizations for improving the results for urban areas. The latter would, however, most likely require changing the urban scheme's parameters, as these schemes have not been developed for African cities.

For deepening the analysis of the different impacts of anthropogenic BC it is crucial to have good observational data, e.g. for BC and particulate matter, vertical profiles of temperature and humidity. Furthermore, long-term measurements

of BC and other aerosols are not only crucial for improving aerosol and chemistry modeling over southern Africa, but also serve for monitoring air quality and assessing air quality management plans and further identifying main sources within the different source categories. So far, only few measurements existed especially of fine particulate matter and BC. However, further monitoring of BC concentrations has recently been initiated: in 2012, continuous BC measurements using the MAAP have commenced at the Secunda, Witbank and Zamdela stations. The network expanded (August 2013) to include additional monitoring stations in the Vaal Triangle operated by the South African Weather Service (SAWS). In addition, aethalometers have been installed in SAWS owned stations in the Karoo and Bojanala areas (September 2014).

Reliable emission inventories with a high temporal and spatial resolution are important in order to improve the modeling of aerosols and air chemistry. More research efforts are needed to create such inventories. For example, further model simulations could aim at analyzing contributions of individual source categories to the modeled BC concentrations. For this, tagging of the individual emitted species would be needed. There is also a need for weekly and diurnal emission profiles as well as the vertical distribution including, for instance, the stack height of important point sources, or the contribution of the different sources within one source category. This is particularly the case for the Pretoria and Johannesburg area, and the industrialized areas in the Mpumalanga Highveld and the Vaal Triangle, where further growth is expected and air pollution is already a recognized problem.

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