Atmospheric processing of iron carried by mineral dust

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Abstract. Nutrification of the open ocean originates mainly from deposited aerosol in which the bio-available iron is likely to be an important factor. The relatively insoluble iron in dust from arid soils becomes more soluble after atmospheric processing and, through its deposition in the ocean, could contribute to marine primary production. To numerically simulate the atmospheric route of iron from desert sources to sinks in the ocean, we developed a regional atmospheric dust-iron model that included parameterization of the transformation of iron to a soluble form caused by dust mineralogy, cloud processes and solar radiation. When compared with field data on the aerosol iron, which were collected during several Atlantic cruises, the results from the higher-resolution simulation experiments showed that the model was capable of reproducing the major observed patterns.

1 Introduction

Mineral dust transported through the atmosphere carries iron-rich minerals that may play a significant role as marine nutrients when deposited into the open ocean. In soil sources, iron is almost insoluble (Zhu et al., 1997), but its solubility can considerably increase during atmospheric transport (e.g., Jickells and Spokes, 2001). The soluble form of iron is considered a critical parameter for marine bioproduction (e.g., Sholkovitz et al., 2012).

Why and how the atmospheric chemical processing of iron evolves remains controversial. There are processes identified as possible explanations for iron solubilization, but their relative importance is not yet well known. It is assumed that longer exposure to solar radiation contributes to faster iron solubilization due to the photochemical reduction of the iron (e.g., Jickells and Spokes, 2001; Hand et al., 2004). Cycling of dust particles in the clouds, in which pH is usually higher than 4, and in the aerosol phase, in which pH is usually substantially lower, can significantly affect iron solubility (Shi et al., 2012). The repeated wetting of the aerosol particle surfaces during their exposure to the cloud conditions should solubilize the iron on the outer surfaces of the particle and enhance the reduction of Fe(III) (Buck et al., 2010). The biomass burning and pollution aerosols are presumed to enhance iron solubility in atmospheric aerosols (mixed dust and combustion aerosol) (e.g., Mahowald et al., 2005; Jickells and Spokes, 2001). Iron in smaller dust particles that are processed in the atmosphere for longer periods has a higher probability of conversion to the soluble form because of physical size sorting (e.g., Baker and Jickells, 2006). Finally, the soil mineralogy plays an important role in iron chemical transformation (Journet et al., 2008).

There are different approaches to parameterize the iron decay rate. Hand et al. (2004) considered that the decay rate coefficient is dependent on the incoming solar radiation and on the cloud cover. Luo et al. (2005) combined the radiation and cloud effects with the influence of the sulfate concentration of polluted air on the iron solubility. In their study, the iron fraction in the dust and the soluble iron fraction in the soils were kept constant. Both these studies specified the decay lifetime as 300 days and assumed that the iron solubility can reach a maximum value of 20% during the two weeks typically required for dust transport across the Atlantic. Moxim et al. (2011) specified that the iron solubilization rate is dependent on the acid chemical processing.
In the mentioned studies, the iron solubility has been estimated, although these methods do not really reproduce the conditions under which the iron is released from aerosol particles once they are deposited onto the ocean surface (where pH is around 8). They show the trends in solubility that might be expected in the behavior of the aerosol in the seawater (Baker et al., 2006). With such methods, being internally consistent, the observed relative changes in the iron solubility could most probably indicate tendencies in the solubility process in seawater (Baker et al., 2006).

The bio-available fraction of the iron deposited on the ocean surface is not well characterized, but it is often assumed to be the soluble fraction (e.g., Mahowald et al., 2005; Sholkovitz et al., 2012). By compiling a large set of global-scale ocean cruise data (G2 in our notation), Sholkovitz et al. (2012) found a consistent hyperbolic dependency between the total aerosol iron and the iron solubility, as a result of the fact that the bulk aerosol is a mixture of the “lithogenic” mineral dust (with origin from natural sources), characterized by high total iron concentration \( T \) and low iron solubility \( s_{%D} \), and the non-lithogenic “combustion” aerosols (from anthropogenic sources), which is characterized by low \( T \) and high \( s_{%D} \). Furthermore, Chen and Siefert (2004) and Baker and Jickells (2006) noted an inverse relationship between the iron solubility in aerosols and the atmospheric dust concentrations.

Baker and Jickells (2006) hypothesized that size reduction of the transported aerosol is a result of higher deposition rate of larger particles due to gravitational settling, and consequently related to decreasing concentration during transport to regions remote from desert sources. In parallel, smaller particles being exposed to chemical processing for a longer time have a larger probability to enhance its iron solubility. Thus, higher solubility for smaller concentrations is likely to be a combined effect of the particle size effects and the acid processing reactions.

Incomplete knowledge of the iron solubilization process in atmospheric dust makes its numerical modeling challenging. Model estimates of the dust deposition in remote marine regions vary by more than a factor of 10 (e.g., Schulz et al., 2012). The current models use relatively coarse horizontal resolutions and therefore describe in less detail the highly variable features of the relevant phenomena, including the emission of the iron minerals from the soils and the mesoscale character of the atmospheric thermodynamics that drives the dust-iron solubilization process. In many modeling studies, simplifications such as the use of a constant iron fraction in the dust emissions or constant iron solubility in the atmosphere are frequently applied (Meskhidze et al., 2003; Mahowald et al., 2009; Shi et al., 2011a).

The objective of this study was to develop a model for simulating the atmospheric processing of iron-containing mineral dust at a regional scale; our focus was on the dust originating from the Sahara. For the first time, high-resolution soil mineralogy was used to calculate the emitted iron fraction.
in the dust. Furthermore, a transport equation for the free iron (the goethite and hematite content) was introduced as one of the model governing equations to parameterize the soluble iron chemical reaction as a function of the mineralogy at the dust sources. We followed here the classification of iron-containing minerals to (i) structural iron embedded in the crystal lattice of aluminosilicates and (ii) iron in oxide/hydroxide called “free-iron” (Lafon et al., 2004).

At this stage, the study did not include aerosols other than the mineral dust, such as aerosols from pollution and biomass burning, which could also contain iron and/or chemically modify the iron carried by dust. The model results were validated against field data collected during several Atlantic cruises.

2 Atmospheric dust-iron model

2.1 Dust and iron emissions and transport

For this study, the DREAM regional dust model (Nickovic et al., 2001; Nickovic, 2005; Perez et al., 2006) was extended by introducing a tracer component for the atmospheric iron. Both the dust and the iron modules are driven online as passive tracers by the NCEP Eta regional atmospheric model (Janjic, 1994 and references therein). The equations for the dust and the total and free iron concentrations are of the same form but differ in their corresponding emissions from their soil sources. The governing equations for the dust concentration and for the total and free iron concentrations include turbulent mixing, vertical and horizontal advection, and wet and dry deposition. The iron chemical reaction is simulated by a first-order equation that was developed as a function of the dust mineralogy, the cloud cover, and the solar radiation. The size distribution of the dust particles is described by eight bins with effective radii of 0.15, 0.25, 0.45, 0.78, 1.3, 2.2, 3.8, and 7.1 μm (Tegen and Lacis, 1996). The first four bins were considered as clay particles and the remaining four as silt particles.

The dynamic part of the DREAM-IRON governing equations is given by

\[ \frac{\partial C_k}{\partial t} + \Delta [C_k] = \left( \frac{\partial C_k}{\partial t} \right)_{\text{SOURCE}} \]  

\[ \frac{\partial T_k}{\partial t} + \Delta [T_k] = \left( \frac{\partial T_k}{\partial t} \right)_{\text{SOURCE}} \]  

\[ \frac{\partial F_k}{\partial t} + \Delta [F_k] = \left( \frac{\partial F_k}{\partial t} \right)_{\text{SOURCE}} \]  

where \( C \) is the dust concentration, \( T \) is the concentration of the total iron in the dust and \( F \) is the concentration of the free iron fraction in the dust. The subscript \( k = 1, 8 \) denotes the particle bins. The operator given below includes the horizontal and vertical advection and the horizontal and vertical diffusion:

\[ \Delta [F_k] = V \cdot \nabla_h \pi_k + (w - w_g) \frac{\partial \pi_k}{\partial z} + \nabla_h \cdot (K_H \nabla_h \pi_k) + \frac{\partial}{\partial z} \left( K_Z \frac{\partial \pi_k}{\partial z} \right) ; \quad \pi_k = (C_k, T_k \text{ or } F_k) \]

Here, \( V = [u,v] \) is the horizontal velocity vector; \( w \) is the vertical velocity; \( w_g \) is the gravitational settling velocity; \( K_H \) and \( K_Z \) are the horizontal and vertical turbulent mixing coefficients; and \( \nabla_h \) is the horizontal nabla operator. The numerical schemes of Nickovic et al. (2001) (including the aerosol emission parameterization) were applied to Eqs. (1)–(3). Equation (3) is introduced for the first time in this type of study to incorporate iron mineralogy into the parameterization of the atmospheric dust-iron model.

2.2 Iron solubility and atmospheric processing

Iron solubility in soils is small, approximately 0.1 % on average (Fung et al., 2000), but it can increase to 80 % as a result of the photo and chemical processing of iron that occurs during the atmospheric dust transport (Mahowald et al., 2005). The iron solubility in dust increases due to clouds, solar radiation and polluted air (Gao et al., 2003; Luo et al., 2005; Jickells and Spokes, 2001; Desboeufs et al., 2001; Hand et al., 2004), but it could also depend on the time that dust spends in the atmosphere (Zhuang et al., 1992; Baker and Jickells, 2006). The atmospheric chemical processing of iron occurs in clouds because they provide a relatively high acidic environment. Observations confirm the relatively high solubility of the iron found in precipitation (e.g., Saydam and Senyuva, 2002). Concerning the solar radiation influence, Zhu et al. (1997) observed higher values of soluble iron during the daytime.

We applied the first-order reaction kinetics approximation to simulate the conversion of iron into its soluble form (e.g., Hand et al., 2004):

\[ \left( \frac{dS}{dt} \right) + K (S - T) = 0. \]  

Here, \( K \) is the decay rate coefficient and \( S \) is the concentration of the soluble iron. Equations (1)–(4) are each applied to the 8 dust particle bins after assuming the same rate of decay for the particles in all of the size bins. Therefore, 32 equations in total are solved at every time step of the model. For further considerations, it is convenient to introduce the following notation: \( t = \frac{t}{f}, s = \frac{s}{f} \), and \( f = \frac{f}{f_0} \).

There are different approaches to parameterize the decay rate coefficient \( K \), as indicated in the Introduction. We assumed that the decay rate coefficient consists of two parts: the first part is associated with processes dependent on the cloud cover and solar radiation \( (K_{CR}) \) and the second part is related to the mineralogy of the dust sources \( (K_M) \):

\[ K = K_{CR} + K_M \]
Similar to Hand et al. (2004), we defined the decay rate due to the cloud and solar radiation effects as

$$K_{CR} = \frac{1}{\tau_{CR}}(\alpha_C + \alpha_R).$$

(6)

Here, $\alpha_C = c$ is the model ratio of cloud cover i.e., fractional cloudiness (e.g., Boers et al., 2010); $\alpha_R = \frac{\langle x_{Temp} \rangle}{\langle x_{Temp} \rangle_{ref}}$, where $\langle \frac{x_{Temp}}{dT} \rangle_{ref}$ is the model temperature tendency caused by solar radiation and $\langle \frac{x_{Temp}}{dT} \rangle = 1 \text{C day}^{-1}$. Note that $\alpha_C$ and $\alpha_R$ are spatiotemporally dependent parameters and $\tau_{CR}$ is the characteristic decay time due to the cloud and radiation effects; the decay time will be specified later.

Following Shi et al. (2011a) we developed a parameterization scheme for $K_M$ that is a function of the iron mineralogy represented by the free-to-total iron ratio. They defined the potential iron solubility as $s_{pot} = \frac{s_{SOIL}}{s_{SOIL, 100 \%}}$, where the soluble fraction, $s_{SOIL}$ was obtained by exposing soil samples to the acidic laboratory conditions for 3 days. As described by Shi et al. (2011a) multiple cycling (3 times) between acidic (pH 2, 24 h) and circumneutral pH (pH 5–6, 24 h) were performed using H$_2$SO$_4$. These pH cycles were made to simulate the cycling of dust between aerosols and clouds during the dust transport.

The soil samples were collected from several African regions known to be dust sources. The sampling locations are marked in Fig. 1 with blue squares. One group of samples was collected from the Sahel region (Mali and three sites in Niger) characterized by high temperatures and relatively high precipitation. These soils were influenced by natural, intense weathering. The second soil group (from Tibesti and Western Sahara) originated from locations where soils are exposed to modest weather conditions. The third soil group (from Bodélé, Tunisia, and two sites in Libya) came from dried paleolake beds exposed to weak chemical weathering.

Shi et al. (2011a) showed that there is a relationship between the degree of chemical weathering (reflected in the free-to-total iron ratio) and $s_{pot}$ for eleven selected sampling sites in Africa:

$$s_{pot} = -22.1 \times f + 15.8(R^2 = 0.44).$$

(7)

The free-to-total iron ratio is an indicator of the soil maturity and the degree of chemical weathering. There was considerable spatial variability among the samples in the free-to-total iron ratio, with generally higher ratios in less weathered soils. The free-to-total iron ratios in the samples ranged from 0.13 (Libya) to 0.6 (Niger). The Sahel soils (Niger sites and Mali), which were mostly weathered, had the lowest $s_{pot}$ values. The weak weathering of the lake sediments (e.g., Bodélé, Tunisia and Libya) resulted in higher $s_{pot}$ values.

To introduce mineralogy as a factor in the iron atmospheric modeling based on the empirical relation in Eq. (7), we developed a technique to map $f$ into the model grid. For that purpose, several geospatial data were employed.

### Table 1. Soil texture classes and their mass percentages of clay and silt.

<table>
<thead>
<tr>
<th>Texture classes</th>
<th>Clay (%)</th>
<th>Silt (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sand</td>
<td>3</td>
<td>5</td>
</tr>
<tr>
<td>Loamy sand</td>
<td>0</td>
<td>19</td>
</tr>
<tr>
<td>Sandy loam</td>
<td>10</td>
<td>26</td>
</tr>
<tr>
<td>Silt loam</td>
<td>12</td>
<td>66</td>
</tr>
<tr>
<td>Silt</td>
<td>5</td>
<td>88</td>
</tr>
<tr>
<td>Loam</td>
<td>18</td>
<td>41</td>
</tr>
<tr>
<td>Sandy clay loam</td>
<td>26</td>
<td>14</td>
</tr>
<tr>
<td>Silty clay loam</td>
<td>33</td>
<td>56</td>
</tr>
<tr>
<td>Clay loam</td>
<td>33</td>
<td>35</td>
</tr>
<tr>
<td>Sandy clay</td>
<td>41</td>
<td>6</td>
</tr>
<tr>
<td>Silty clay</td>
<td>46</td>
<td>47</td>
</tr>
<tr>
<td>Clay</td>
<td>61</td>
<td>19</td>
</tr>
</tbody>
</table>

First, we calculated the iron percentages in iron-carrying minerals (illite, kaolinite, smectite, iron oxides and feldspars) from the GMINER30 global 1 km database (Nickovic et al., 2012). These percentages were specified for dust sources assumed to exist for the following 1 km USGS land cover types: low sparse grassland, bare desert, semi desert, sand desert, semi desert shrubs, and semi desert sage (Nickovic et al., 2001). The dust sources for latitudes higher than 55° were defined as nonproductive dust soils. The emitted effective mineral fractions were calculated as the weighted means with respect to the clay and silt soil contents. The clay and silt values were obtained from the gridded soil texture classes in the STATSGO-FAO database in which the FAO 5 min global soil texture is remapped onto a global 30 s (~1 km) grid (USDA, 1994). Table 1 shows a correspondence between the soil texture classes and the mass percentages of clay and silt as used in the model. The values in the table were estimated from the textural triangle by combining data from Tegen et al. (2002) and Shirazi et al. (2001). The sand was ignored because it does not contribute to the longer-range dust-iron transport addressed by our study. Therefore, we assumed the emitted dust in the model to be a mixture of clay and silt particles only.

Over the North African dust sources, the total iron content in the different types of iron-containing minerals varies substantially. Table 2 shows the percentages of the total iron in the different minerals that were used in the model, and the ranges reported in the literature (Thomson and Hower, 1975; Journet et al., 2008; Paris et al., 2011). Although there are differences in the chemical reactivity of the goethite and the hematite, these minerals were here considered to be iron oxides (e.g., Claquin et al., 1999; Shi et al., 2011a).

### 2.3 Total iron distribution in soil and decay rates

By combining the Table 2 data and the mineral fractions from GMINER30, a 1 km grid dataset for the total iron...
Table 2. Content of total iron in different minerals as used in the model and its range based on values reported in Thomson and Hower, 1975; Journet et al., 2008; Paris et al., 2011.

<table>
<thead>
<tr>
<th>Mineral</th>
<th>Illite</th>
<th>Kaolinite</th>
<th>Smectite</th>
<th>Goethite and Hematite (clay and silt)</th>
<th>Feldspar</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total Fe (%) used in DREAM</td>
<td>4.8</td>
<td>0.7</td>
<td>16.4</td>
<td>66</td>
<td>2.5</td>
</tr>
<tr>
<td>Total Fe (%) from literature</td>
<td>3.38–4.65</td>
<td>~0.7</td>
<td>2.55–23</td>
<td>57.5–77.4</td>
<td>0.13–0.54</td>
</tr>
</tbody>
</table>

Fig. 1. Geographic distribution of iron-related soil parameters over northern Africa. (a) Total iron; (b) Free-to-total iron ratio $f$; (c) Potential solubility $s_{pot}$.

fraction, $t_{SOIL} \times 100\%$, in the clay-silt mixture was generated. As shown in Fig. 1a, the total iron demonstrated significant geographical variability. In most of the Sahara, the total iron ranged from 2–3.6 %, except for some parts of northern Egypt, Libya and Algeria, where the values reached 6 % or more. The total iron amounts were generally higher in the Sahelian region. The GMINER30 values were within the range of the published observational evidence. Guieu et al. (2002) showed that the average observed total iron originating from Tunisia, Morocco, Niger and Algeria is 4.45 %. Lafon et al. (2004) reported that the total iron ranges from 6.2 to 8.7 % in Saharan and Sahelian samples. Other observations indicate that the total iron in samples collected in the
Sahara and the Sahel is in the range of 0.8–12.2 % (Shi et al., 2011a).

Figure 1b displays \( f \) values mapped from GMINER30; the locations of the observation sites are marked with the blue squares. The mapping indicated that there is a belt in the Sahel with rather high \( f \) values up to \( \sim 0.6 \); that North Africa, parts of Algeria, Libya and Egypt are characterized by \( f \) values ranging from 0.4 to 0.45; and that the \( f \) values in the remaining Saharan region varied from less than 0.1 to 0.25.

Figure 2 is the scatter diagram of the \( f \) values observed at 11 sites and of the values interpolated from GMINER30 in the same locations, with strong correlation achieved between the observed and the interpolated values \( (R^2 = 0.76) \). This correlation provides indirect confirmation that the GMINER30 data compare well with observations. The gridded \( s_{\text{pot}} \) calculated from Eq. (7) is shown in Fig. 1c. These calculations indicated that, in general, the soils of the central Saharan belt are sources of potentially more reactive iron characterized by low free-to-total iron ratios.

The gridded values of \( f \) and \( s_{\text{pot}} \) were used to parameterize the iron atmospheric processing as a function of the mineralogy at the soil sources. To specify the decay rate \( K_M \), we integrated with respect to time from \( t = 0 \) to \( t_{\text{pot}} \) the part of Eq. (4) addressed to the mineralogy:

\[
\ln\left(100 s_{\text{pot}}\right) = \ln\left(100 s_0\right) - K_M t_{\text{pot}},
\]

where we assumed \( s_0 = 0.1 \% \) to be a typical solubility in the soil sources (e.g., Moxim et al., 2011). \( t_{\text{pot}} = 75 \) days has been chosen as a typical time needed for the solubility to achieve \( s_{\text{pot}} \) in the atmosphere. It is within the range of values reported by other studies cited in the Introduction and it is tuned to reduce differences between observations and model simulations.

After neglecting \( s_0 \) as a small value compared to 100 %, we evaluated the decay rate due to the mineralogy as

\[
K_M = \frac{1}{t_{\text{pot}}} \ln\left(1 - \frac{s_{\text{pot}}}{100}\right).
\]

3 Simulation experiments

A set of model experiments was conducted to simulate the iron atmospheric cycle for various oceanographic cruises in the Atlantic. Two groups of aerosol observations were studied. One group (hereafter called G1) relates to parts of Atlantic cruises JCR, BUCK, ANT23-1, PEL, M55 (1) and M55 (2) (Baker et al., 2003; Buck et al., 2010) in which aerosol is likely to be from the Saharan dust sources. The second group of observations (G2) represents a compilation of aerosol samples collected during various Atlantic cruises on both dusty and nondusty days (Sholkovitz et al., 2012; data shown in their Figs. 5 and 6). These observations were used to investigate how our simulations compare with the observed hyperbolic dependency between the iron solubility and the total iron loading. Table 3 summarizes the dates and the start and end points of G1 cruises for which we interpolated the model parameters, while in Table 4 are given regions of G2 cruises and their references.

For JCR, BUCK and ANT23-1 with paths closer to the African coastline, we used a smaller model domain with borders 36° W, 22° E, 3° S and 37° N. For PEL, M55 (1) and M55 (2), the model domain was extended to 59.5° W, 29.5° E, 1.5° S and 33.5° N, in order to cover the longitudinal path of M55 (1). For both setups, a horizontal model resolution of 0.25 deg was applied.

For each cruise experiment, the meteorological initial and boundary conditions for the atmospheric model driver were updated daily from the gridded 0.5-deg ECMWF objective analyses archive. The simulated 24 h dust and iron concentrations from a previous day were used as the initial conditions for a current day. The experiments were initiated with

\[\text{Table 3. Group G1 of cruises from which data have been used in this study.}\]

<table>
<thead>
<tr>
<th>Cruise name</th>
<th>Start and end date</th>
<th>Start and end latitude; longitude</th>
</tr>
</thead>
<tbody>
<tr>
<td>JCR\textsuperscript{b,c}</td>
<td>17 to 19 Sep 2001</td>
<td>(22.7; −18.3) to (30.8; −20)</td>
</tr>
<tr>
<td>PEL\textsuperscript{d}</td>
<td>5 to 7 Oct 2002</td>
<td>(33.5; −22.7) to (30.8; −20)</td>
</tr>
<tr>
<td>M55 (1)\textsuperscript{f}</td>
<td>15 to 26 Oct 2002</td>
<td>(10.8; −56.2) to (8.5; 6.6)</td>
</tr>
<tr>
<td>M55 (2)\textsuperscript{f}</td>
<td>2 to 9 Nov 2002</td>
<td>(9.5; −24.7) to (8.2; −18)</td>
</tr>
<tr>
<td>BUCK\textsuperscript{f}</td>
<td>21 to 26 Jul 2003</td>
<td>(23; −27.3) to (15; −29)</td>
</tr>
<tr>
<td>ANT23-1\textsuperscript{g}</td>
<td>1 to 2 Nov 2005</td>
<td>(12.3; −20.5) to (8.7; −18.8)</td>
</tr>
</tbody>
</table>

\textsuperscript{a} Only days dominated by dust (according to A. R. Baker, personal communication, 2012 and Buck et al. (2010)) are included. \textsuperscript{b} Baker et al. (2003); \textsuperscript{c} Baker et al. (2006); \textsuperscript{d} Buck et al. (2010) ; \textsuperscript{f} Baker et al. (2010).
a zero-concentration “cold start” four days before a starting cruise date to remove model spin-up effects.

4 Model evaluation

There are significant sources of errors in both modeling and observations. In the dust model intercomparison studies of Uno et al. (2006) and Todd et al. (2008) the predicted surface level concentrations sometimes show differences of at least an order of magnitude. On the observational side, aerosol concentrations in general and in particular iron fraction and/or its soluble part are difficult to measure accurately at low concentrations observed (Mahowald et al., 2009; Baker and Croot, 2008). Additional uncertainty in observations is originating from the fact that the total iron was often assessed assuming its fraction in dust mass concentration to be 3.5 % (e.g., Baker and Croot, 2008; Mahowald et al., 2009; Sholkovitz et al., 2012).

Therefore, when accuracy of dust-iron models is assessed, one has to draw conclusions with caution. Yet, there is a necessity to understand modeling performance in direct comparison against observations, because large gaps in iron-related observations could be at least partly compensated by information from model simulations. Several studies report on direct point-to-point model-observations comparisons (Luo et al., 2005, 2008; Hand et al., 2004; Mahowald et al., 2009). Based on several model studies used to extrapolate observation data, Mahowald et al. (2009) concluded that annual averaged model iron surface concentrations and daily averaged observations differ by a factor of 50–1000 %. On the other hand, their study shows that in direct point-to-point comparison, scattering among the individual cruises is much smaller than for the annual averages (with most of the points falling within ± one order of magnitude interval). Their results also indicate that the model overpredicts iron especially in places with very low concentrations and depositions.

Following the mentioned approach of direct comparison of simulated and observed parameters along the cruises, we show performance of the near-surface model total iron concentration $T$, the soluble iron concentration $S$ and the iron solubility ($S_{%}$). Figure 3 shows $T$, $S$, $S_{%}$ simulated for M55 (1) when the dust was the aerosol likely to originate from Saharan dust sources. Namely, datasets ICR, ANT23-1, PEL, M55 (1) and M55 (2) include only samples which had visible Saharan dust on the filters and should therefore have dust as a major component of the total aerosol (A. Baker, personal communication, 2012). Similarly, BUCK data include only samples marked as Saharan for the air mass regime (Table 1 in Buck et al., 2010).

The cruise track and the markers indicating the positions of the cruise vessel at selected model valid times are plotted as a reference to the simulated fields.

In Fig. 3, we plotted maps valid for 12:00 UTC on 17, 21 and 25 October to illustrate how much the model variables fluctuate spatially and temporally. Only the variables over the ocean are shown. Note that the trade winds produced a typical east–west shape of the fields, although on 17 October, there was a more north–south weave-like variability of the fields. As a result of using relatively fine resolutions for the mineral sources and for the driving atmospheric model, the total iron concentration field $T$ (Fig. 3a–c) is characterized by many mesoscale structures. The patterns of the dust concentration $C$ (not shown) are similar in shape to those of $T$. Some similarities in shape between the total concentration $T$ and the soluble iron concentration $S$ are also evident (Fig. 3d–f); however, patterns such as the local maxima in the two fields are not comparable because $S$ additionally varies due to the chemical transformation. Such differences in $T$ and $S$ produce a highly variable structure of their derived field, the solubility $S_{%}$ (Fig. 3g–i). Considerable daily variability in $S_{%}$ is also observed, with the value sometimes changing several times between consecutive days (e.g., Baker et al., 2003, 2010; Baker and Jickells, 2006; Buck et al., 2010; Sholkovitz et al., 2012).

Table 4. Group G2 of cruises from which data have been used in this study.

<table>
<thead>
<tr>
<th>Cruise reference number as in Sholkovitz et al. (2012)</th>
<th>Reference</th>
<th>Region</th>
</tr>
</thead>
<tbody>
<tr>
<td>10</td>
<td>Wu et al. (2007)</td>
<td>western North Atlantic (Sargasso Sea)</td>
</tr>
<tr>
<td>12</td>
<td>Chen (2004)</td>
<td>Atlantic</td>
</tr>
<tr>
<td>15</td>
<td>Buck et al. (2010)</td>
<td>eastern North Atlantic</td>
</tr>
<tr>
<td>17</td>
<td>Baker et al. (2006)</td>
<td>North Atlantic</td>
</tr>
<tr>
<td>18</td>
<td>Chen and Siebert (2004)</td>
<td>North Atlantic</td>
</tr>
<tr>
<td>19</td>
<td>Johansen et al. (2000)</td>
<td>Tropical North Atlantic</td>
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<td>21</td>
<td>Sedwick et al. (2007)</td>
<td>North Atlantic (Sargasso Sea)</td>
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<tr>
<td>22</td>
<td>Church, Sedwick and Sholkovitz, unpublished data</td>
<td>Atlantic (Bermuda tower samples)</td>
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<td>23</td>
<td>Zhu et al. (1997)</td>
<td>Atlantic (Barbados)</td>
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Fig. 3. Spatial and temporal features of simulated dust-related fields. (a–c): the instantaneous total iron concentration in dust $T$ (ng m$^{-3}$) at 12:00 UTC for 17, 21 and 25 October 2002; (d–f): the same as in (a) but for the soluble iron concentration $S$ (pg m$^{-3}$); (g–i): the same as in (a) but for the iron solubility $s$%.

Next, we focus on comparing the model results with the observations from the daily samples of the $C$, $T$ and $S$ taken during the G1 cruises, which provided a total of 32 identified days when dust was a dominating aerosol.

Our experiments showed that for most samples, the model dust concentration is within an order of magnitude of accuracy (Fig. 4a), with values for the bias and the root mean square error of 1 and 40 µg m$^{-3}$, respectively. The scatter diagram of the same values is shown in Fig. 4b. For a single...
day of 6 November 2002 (M55 (2) cruise), there was an exceptionally large error (predicted \( C = 2 \mu g m^{-3} \) vs. observed \( C = 104 \mu g m^{-3} \)), which caused large underestimates in the predictions of \( T \) and \( S \) for that particular day, as shown below.

The simulated total iron concentrations should in general deviate more from observations because of the uncertainties added from the specification of the iron sources. In our simulations, most of the predicted and observed daily \( T \) values differed by no more than one order of magnitude (Fig. 4c), although the model generally had a tendency to underpredict \( T \) (bias = \(-414 ng m^{-3} \); RMSE = \( 1171 ng m^{-3} \)). When compared with the results of Mahowald et al. (2008) for the same range of values (10–600 ng m\(^{-3}\)) our experiments show less scattering, possibly because our simulations have much higher model resolution and include more detailed specification of iron sources in desert soils.

The prediction of the soluble iron concentration \( S \) (Fig. 4e) is more uncertain because parameterizations of the solubility process are still based on insufficient information about the iron chemical transformations. As a result, in our experiment approximately 25 % of the daily model predictions deviated from the observed daily values by one to two orders of magnitude, leading to negative bias and RMSE values being larger than those of the \( C \) and the \( T \) (bias = \(-7466 pg m^{-3} \); RMSE = 24 771 pg m\(^{-3}\)).

In Fig. 5, we show the soluble and total iron values from the Atlantic cruises in Sholkovitz et al. (2012) (shown in their Fig. 5b) combined with our model data interpolated along the path of the G1 cruises (note that the G2 dataset in our analysis also include most of the G1 data). Most of the model values reproduced the observed hyperbolic trend, which justifies the suitability of the parameterization methods developed in this study.
Figure 5. Fe solubility versus total Fe for the Atlantic sites/cruises. The plotted two groups of data: blue diamonds are from datasets 15, 16, 17, 18, 21, 22, 23 and unpublished data of Powell and Baker, as shown in Sholkovitz et al. (2012; Figure 5B); pink squares are simulated values.

In Fig. 6, we plot together their data and our simulated values. For observations where the solubility gets higher values, the samples had spent at least five days over the ocean before collection, thus aerosol was exposed to relative longer atmospheric processing (see A. Baker description at http://www.uea.ac.uk/~e780/solubility.htm). The model values are concentrated in the lower right quadrant (higher C and lower s%) and in general follow the observed trend for higher concentrations. Actually, the model values group mainly around observations of aerosol originating from Sahara because in our parameterization the influence of other aerosols originating from anthropogenic sources is neglected. Ito (2013) however showed that his model which includes iron-containing aerosols from shipboard pollution was able to reproduce the highly soluble part of the hyperbolic trend, demonstrating so the importance of including the anthropogenic aerosol.

Although the model is applied over Northern Africa (Sahara, Sahel) and subtropical North Atlantic, our parameterization of the iron faith in the atmosphere is not specifically addressed to any particular geographic region. Certainly, the atmospheric driving forces for the iron dissolution are specific for a selected geographic domain such as much less clouds coverage than in the equatorial and mid-latitude areas (and consequently less influence on iron), or a specific incident angle of the solar radiation to the earth surface in the subtropics.

In order to explore how in our simulations the iron processing reacts to different effects, we evaluated a set of sensitivity experiments for the M55 cruise by executing the model without clouds, with constant Fe in sources and without radiation, and then compared the results with those of the reference model containing all three effects acting together (Fig. 7). The shown graph indicates that the impact of clouds is marginal due to prevailing cloud-free conditions. With the constant 3.5% iron in the sources used, the soluble values drop down by about 20% or less. Finally, with no radiation effects included, the soluble iron reduces for about 40%.

5 Summary

In this study, we extended a mineral dust atmospheric model by adding a component for the atmospheric cycle of iron carried by dust. Unlike studies using global models with coarse resolutions, we performed experiments using a regional model with a horizontal resolution fine enough to resolve the mesoscale atmospheric dynamics and to more accurately describe the emission, transport and deposition of the dust and iron. The iron fractions in desert sources were estimated using detailed geographic distributions of the major soil minerals. This approach considerably improves the
usual practice of setting a constant value for the iron content in soils. Furthermore, we developed a parameterization method that defines the decay rate coefficient for iron reduction as a function of the dust mineralogy, cloud processing and solar radiation. The rate coefficient was designed to vary in space and time during the model execution.

Results from the dust-iron modeling experiments were compared with observations collected along parts of several Atlantic Ocean cruise routes dominated by dust aerosol. The predicted dust-iron parameters, especially the iron solubility, showed high temporal and spatial variability. The model daily concentrations of the dust aerosol and of the total and soluble iron interpolated along the cruise routes generally showed good agreement with the sampled data. However, the total and the soluble iron were underestimated to some degree, possibly because of eventual smaller iron emissions at the soil origins. We also demonstrated that the simulated relationship between the solubility and the total iron amount is generally comparable to the observed hyperbolic trend for lower concentrations. This study was solely focused on desert dust because we neglected the influence of other aerosols originating from anthropogenic, biomass burning and volcanic sources that could contribute to the atmospheric iron cycle (Jickells and Spokes, 2001).

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Fig. 7. Ratios of iron soluble concentrations without/with clouds (blue dots), without/with variable mineralogy (orange dots), and without/with radiation (red dots) included.


