The contrasting roles of water and dust in controlling daily variations in radiative heating of the summertime Saharan heat low

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The contrasting roles of water and dust in controlling daily variations in radiative heating of the summertime Saharan heat low

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Abstract. The summertime Sahara heat low (SHL) is a key component of the West African monsoon (WAM) system. Considerable uncertainty remains over the relative roles of water vapour and dust aerosols in controlling the radiation budget over the Sahara and therefore our ability to explain variability and trends in the SHL, and in turn, the WAM. Here, new observations from Fennec supersite-1 in the central Sahara during June 2011 and June 2012, together with satellite retrievals from GERB, are used to quantify how total column water vapour (TCWV) and dust aerosols (from aerosol optical depth, AOD) control day-to-day variations in energy balance in both observations and ECWMF reanalyses (ERA-I). The data show that the earth-atmosphere system is radiatively heated in June 2011 and 2012. Although the empirical analysis of observational data cannot completely disentangle the roles of water vapour, clouds and dust, the analysis demonstrates that TCWV provides a far stronger control on TOA net radiation, and so the net heating of the earth-atmosphere system, than AOD does. In contrast, variations in dust provide a much stronger control on surface heating, but the decreased surface heating associated with dust is largely compensated by increased atmospheric heating, and so dust control on net TOA radiation is weak. Dust and TCWV are both important for direct atmospheric heating. ERA-I, which assimilated radiosondes from the Fennec campaign, captures the control of TOA net flux by TCWV, with a positive correlation \((r = 0.6)\) between observed and modelled TOA net radiation, despite the use of a monthly dust climatology in ERA-I that cannot capture the daily variations in dustiness. Variations in surface net radiation, and so the vertical profile of radiative heating, are not captured in ERA-I, since it does not capture variations in dust. Results show that ventilation of the SHL by cool moist air leads to a radiative warming, stabilising the SHL with respect to such perturbations. It is known that models struggle to capture the advective moistening of the SHL, especially that associated with mesoscale convective systems. Our results show that the typical model errors in Saharan water vapour will lead to substantial errors in the modelled TOA energy balance (tens of W m\(^{-2}\)), which will lead to errors in both the SHL and the WAM.

1 Introduction

The Sahara lies under the descending branch of the Hadley circulation and during summer the intense solar heating combined with the arid environment leads to large sensible surface heat fluxes and the formation the Saharan heat low (SHL). This increases the pressure gradient from the Gulf of Guinea to the Sahara, driving the West African monsoon (WAM); and variations in the SHL modify the WAM on timescales from days to decades (Thorncroft and Blackburn, 1999; Sultan and Janicot, 2003; Parker et al., 2005; Peyrillé and Lafore, 2007; Biasutti et al., 2009; Lavaysse et al., 2009, 2010; Chauvin et al., 2010; Xue et al., 2010; Martin and Thorncroft, 2014; Martin et al., 2014; Dong and Sut-
Similarly ventilation of the Sahara by the Atlantic Inflow involves mesoscale flows that are a challenge for global models (Grams et al., 2010; Todd et al., 2013). Since clouds, water vapour and dust are all important to the Sahara’s radiative energy balance, such model errors in convection, clouds, haboobs and advection of water vapour will all affect modelled radiative energy balances and hence climate.

There is a clear need to establish the controls on the radiation budget over the Sahara and evaluate models. In this paper we use observations of surface radiative fluxes from Fennec supersite-1 in the central Sahara and retrievals of TOA fluxes from satellite data to investigate how dust and water vapour interact. Results in Sect. 3 show that TCWV (total column water vapour) and AOD are correlated and we cannot completely isolate the effects of either TCWV or dust. However, TCWV and AOD have sufficiently independent variations, and sufficiently distinct impacts at solar and infrared wavelengths, which conform with physical principles, that the results give unique insights into their contrasting roles in the central Sahara. Section 2 describes methods, Sect. 3 presents results, Sect. 4 contains discussion and conclusions are in Sect. 5.

2 Method

We use data from Fennec supersite-1 in the central Sahara, located at Bordj-Badji Mokhtar (BBM) at 21.4°N 0.9°E (in the very south of Algeria, close to the triple point of Algeria, Mali and Niger), close to the SHL’s climatological centre and the dust maximum (Marsham et al., 2013a), together with corresponding values from ERA-Interim (Dee et al., 2011), and satellite TOA fluxes (Harries et al., 2005; Dewitte et al., 2008). These satellite TOA fluxes are produced using a narrow to broad band conversion of SEVIRI (Spinning Enhanced Visible and Infrared Imager) radiance measurements. These are scaled by co-located GERB (Geostationary Earth Radiation Budget experiment) measurements and converted into broadband fluxes measurements at a horizontal resolution of $3 \times 3$ SEVIRI pixels (0.32–4 microns in the shortwave, and 4–100 microns in the longwave). This enhancement gives a spatial resolution of 9 km at nadir, compared with the 45 km of the native GERB. These observed fluxes are from clear and cloudy skies, but we also use the European Organisation for the Exploitation of Meteorological Satellites’ MPEF (Meteorological Product Extraction Facility) cloud mask as a simple measure of cloud cover. ERA-Interim uses aerosol climatologies so it cannot capture day-to-day variations in dust. Radiosonde data from the Fennec supersite were assimilated into ERA-I, which will have improved its representation of the thermodynamic profile (see Garcia-Carreras et al., 2013, for impacts of assimilation of Fennec radiosondes on the Met Office global forecast model).
Fennec data are from intensive observation periods (IOPs) in June 2011 and 2012, when a Cimel sun photometer provided aerosol optical depths (AODs) at 675 nm, a Kipp & Zonen radiometer mounted at 2 m provided measurements of broad-band radiative fluxes and 3- to 6-hourly radiosonde observations were available (Marsham et al., 2013a). The sun-photometer is part of the AERONET program (Holben et al., 1998) and cloud-screened AOD retrievals are only available during the day. Level-2 AOD data are not available for 2012 since not all data meet level-2 requirements. However, the 0.675 nm AODs are still reliable. We therefore use level-2 data for 2011 and level-1.5 for 2012, noting that using only 2011 data does not affect our conclusions. We use the radiosondes to compute column water vapour from the surface to 300 hPa (a height consistently reached by the radiosondes), which we refer to simply as “total column water vapour (TCWV)”. During June 2011 BBM was regularly cooled by nocturnal monsoon flows and embedded cold pools giving substantial variability in TCWV (Marsham et al., 2013a), and qualitatively similar weather events were observed during June 2012.

In order to study the day-to-day variations in the energy budget, we average all data to their daily means. Complete surface flux data were only available for 11 days in June 2011 (9, 10, 18–20, 23–27, 30 June) and 25 days in June 2012 (all except 4, 16–18, 30 June). Some dates had short data gaps (around 2 hours on 2 days, but otherwise an hour or less) and these gaps were interpolated across in order to include 7, 17, 21, 22 June 2011 and 16–17 June 2012. This gave an improved range of AODs, albeit with increased uncertainties in surface fluxes. Fluxes from these days with some interpolation are marked by squares in Figs. 1 to 4, and the effects of interpolation are discussed in Sect. 3, where it is seen that results from these days are physically consistent with other data from days without interpolation. The surface flux data from Kipp and Zonen radiometers have slightly different spectral ranges to the satellite-borne GERB: Kipp and Zonen are 0.3 to 2.8 μm in the shortwave and 4.5 to 42 μm in the longwave, whilst GERB is 0.32 to 4 μm and 4 to 100 μm. This means that the surface-based Kipp and Zonen can miss up to 3.5 W m⁻² net shortwave atmospheric heating as would be seen by GERB and up to 3.8 W m⁻² of the net longwave as would be seen by GERB (Banks et al., 2014). This introduces errors of up to 4 W m⁻² in our inferred direct atmospheric radiative heating rates, but does not affect our analysis and conclusions, which is focused on the controls on the variability of these rates, rather than their absolute values.

Sun-photometer AODs were available from 8 June 2011 (with no observations on 13 June) and 1 to 28 June 2012 (with no observations on 17–19 June). Radiosondes were available from 8 June 2011 and 1 to 26 June 2012. The number of observations contributing to the daily mean is variable for AODs, since observations are only made when it is cloud free, but all days except one had at least eight AOD observations and the daily mean AOD range of 0.2 to 2.7 is similar to that of the observation range in AODs (0.2 to 3.9) and the diurnal cycle in AOD is weak (Marsham et al., 2013a; Banks et al., 2014). Overall this gave 36 days with surface data, observed AOD and observed TCWV and 44 days with TCWV and AOD.

3 Results

In order to determine how the changing amounts of water and dust over BBM affect the changing radiative heating at the surface, TOA and within the atmosphere we analyse relationships between the daily means of key variables, using both observed quantities and the equivalent from ERA-I at the location of BBM. ERA-I uses a climatological AOD field and so cannot capture the observed daily variability in AOD. This, in effect, represents a quasi-control experiment for dust variability. Net fluxes are defined as downward, with increased net downward flux corresponding to increased shortwave heating or reduced longwave cooling. All correlations and slopes of linear regression lines discussed are listed in Table 1 (correlations in bold are significant at the 90 % level). Relationships are shown using days where surface flux data are available (referred to as “Good surface data”), and for all available data (“All data”) where surface flux data are not required. The regressions are very similar whichever data set is used and values in the text are for “Good surface data”, unless otherwise noted. The effect of subsampling is small for ERA-I, showing that general lessons can be drawn from the observational data, despite the limited time-span of the data set. Similarly, for both June 2011 and 2012 analysed water vapour at 850 and 925 hPa and AODs from MISR, Deep blue (Terra and Aqua) and OMI are all within 1 standard deviation of their mean values (not shown) and there is no indication that the weather regimes affecting BBM in these periods were anomalous.
Table 1. Gradients of best-fit straight lines (i.e. regression coefficients) for listed relationships, values in [ ] are normalised by standard deviation of TCWV or AOD. Values in ( ) are correlation coefficients (bold values are significant at 90 % level). For ERA-I observed AODs are used. Standard deviation in TCWV in ERA-I = 4.7 kg m\(^{-2}\) (4.5 for “All data”). For observations 4.7 kg m\(^{-2}\) (4.4 for “All data”). Standard deviation in AOD for observations is 0.68 (0.65 in “All data”).

<table>
<thead>
<tr>
<th></th>
<th>Good surface data</th>
<th>All data</th>
<th>Good surface data</th>
<th>All data</th>
</tr>
</thead>
<tbody>
<tr>
<td>TCWV : AOD (kg m(^{-2}))</td>
<td>0.04 (0.29)</td>
<td>0.04 (0.30)</td>
<td>0.02 (0.16)</td>
<td>0.02 (0.14)</td>
</tr>
<tr>
<td>AOD : TOA Net (W m(^{-2}))</td>
<td>5.3 [3.6] (0.26)</td>
<td>3.5 [2.3] (0.17)</td>
<td>1.7 (0.12)</td>
<td>0.33 (0.02)</td>
</tr>
<tr>
<td>AOD : TOA Net LW (W m(^{-2}))</td>
<td>10.5 [7.2] (0.33)</td>
<td>8.5 [5.5] (0.26)</td>
<td>2.4 (0.12)</td>
<td>1.0 (0.05)</td>
</tr>
<tr>
<td>AOD : TOA Net SW (W m(^{-2}))</td>
<td>−5.2 [−3.6] (−0.28)</td>
<td>−5.0 [−3.3] (−0.26)</td>
<td>−0.72 (−0.07)</td>
<td>−0.70 (−0.06)</td>
</tr>
<tr>
<td>TCWV : TOA Net (W kg(^{-1}))</td>
<td>2.2 [10.4] (0.74)</td>
<td>2.1 [9.2] (0.68)</td>
<td>1.3 (0.66)</td>
<td>1.4 (0.66)</td>
</tr>
<tr>
<td>TCWV : TOA Net LW (W m(^{-2}))</td>
<td>3.2 [15.0] (0.68)</td>
<td>3.0 [13.3] (0.63)</td>
<td>1.8 (0.61)</td>
<td>2.0 (0.65)</td>
</tr>
<tr>
<td>TCWV : TOA Net SW (W kg(^{-1}))</td>
<td>−0.98 [−4.6] (−0.36)</td>
<td>−0.9 [−4.1] (−0.33)</td>
<td>−0.48 (−0.30)</td>
<td>−0.49 (−0.30)</td>
</tr>
<tr>
<td>TOA Net SW : TOA Net LW</td>
<td>−1.38 (−0.80)</td>
<td>−1.35 (−0.79)</td>
<td>−1.44 (−0.78)</td>
<td>−1.39 (−0.72)</td>
</tr>
<tr>
<td>AOD : Surface Net (W m(^{-2}))</td>
<td>−13.1 [−9.0] (−0.70)</td>
<td>NA</td>
<td>3.4 (0.34)</td>
<td>3.2 (0.31)</td>
</tr>
<tr>
<td>AOD : Surface Net SW (W kg(^{-1}))</td>
<td>−31.9 [−21.8] (−0.87)</td>
<td>NA</td>
<td>−1.6 (−0.12)</td>
<td>−1.7 (−0.12)</td>
</tr>
<tr>
<td>AOD : Surface Net LW (W m(^{-2}))</td>
<td>20.7 [14.2] (0.81)</td>
<td>NA</td>
<td>5.0 (0.28)</td>
<td>4.9 (0.27)</td>
</tr>
<tr>
<td>TCWV : Surface Net (W kg(^{-1}))</td>
<td>−0.20 [−0.96] (−0.07)</td>
<td>NA</td>
<td>0.76 (0.53)</td>
<td>0.85 (0.57)</td>
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<tr>
<td>TCWV : Surface Net LW (W kg(^{-1}))</td>
<td>2.0 [9.3] (0.54)</td>
<td>NA</td>
<td>2.2 (0.84)</td>
<td>2.2 (0.85)</td>
</tr>
<tr>
<td>TCWV : Surface Net SW (W kg(^{-1}))</td>
<td>−1.8 [−8.2] (−0.33)</td>
<td>NA</td>
<td>−1.4 (−0.69)</td>
<td>−1.4 (−0.68)</td>
</tr>
<tr>
<td>Surface Net SW : Surface Net LW</td>
<td>−0.61 [−0.88]</td>
<td>NA</td>
<td>−1.1 (−0.83)</td>
<td>−1.1 (−0.82)</td>
</tr>
<tr>
<td>AOD : Atmospheric Net (W m(^{-2}))</td>
<td>18.5 [12.1] (0.62)</td>
<td>NA</td>
<td>−1.75 (−0.13)</td>
<td>−2.9 (0.21)</td>
</tr>
<tr>
<td>AOD : Atmospheric Net LW (W m(^{-2}))</td>
<td>−10.2 [−6.7] (−0.41)</td>
<td>NA</td>
<td>−2.65 (−0.18)</td>
<td>−3.9 (−0.26)</td>
</tr>
<tr>
<td>AOD : Atmospheric Net SW (W kg(^{-1}))</td>
<td>26.7 [17.5] (0.93)</td>
<td>NA</td>
<td>0.91 (0.13)</td>
<td>1.0 (0.15)</td>
</tr>
<tr>
<td>TCWV : Atmospheric Net (W kg(^{-1}))</td>
<td>2.4 [10.7] (0.56)</td>
<td>NA</td>
<td>0.51 (0.26)</td>
<td>0.54 (0.27)</td>
</tr>
<tr>
<td>TCWV : Atmospheric Net LW (W kg(^{-1}))</td>
<td>1.2 [5.4] (0.34)</td>
<td>NA</td>
<td>−0.41 (−0.20)</td>
<td>−0.36 (−0.17)</td>
</tr>
<tr>
<td>TCWV : Atmospheric Net SW (W kg(^{-1}))</td>
<td>0.78 [3.4] (0.19)</td>
<td>NA</td>
<td>0.91 (0.93)</td>
<td>0.90 (0.91)</td>
</tr>
<tr>
<td>Atmos Net SW : Atmos Net LW</td>
<td>−0.39 [−0.45]</td>
<td>NA</td>
<td>−0.73 (−0.35)</td>
<td>−0.79 (−0.36)</td>
</tr>
</tbody>
</table>

There are correlations between dust and water (discussed below) which mean that effects of either cannot be completely isolated from the other, but nevertheless the approach allows identification of how variations in these variables affect radiative heating. Figure 1 shows that there is a significant tendency for more dust with more TCWV, although there are a few dry dusty days (correlation = 0.29). The use of surface flux data with some interpolation (shown by squares) allows study of more days with high AODs. The mechanisms underlying this correlation are understood: Marsham et al. (2013a) shows how moist monsoon surges from the south are associated with dust at BBM. This is because the moist surges are associated with both dusty haboobs and moist nocturnal low-level jets (LLJs) that together dominate the dust uplift at BBM in June 2011 (Marsham et al., 2013a; Allen et al., 2013). The association between dust and water vapour is consistent with Fig. 16 in Marsham et al. (2013a), which shows a statistical link between AOD and cloud cover at BBM. Intense dust uplift does sometimes occur in dry air, however, mainly in the dry Harmattan LLJs (Marsham et al., 2013a; Allen et al., 2013).

3.1 Control of TOA net radiation by water (TCWV) and aerosols (AOD)

Daily mean net TOA radiation is always positive (i.e. downwards) and has a mean value of 26 W m\(^{-2}\), i.e. there is warming of the earth-atmosphere system throughout the period (Fig. 2a). Net heating varies between around 0 and 70 W m\(^{-2}\), or approximately 0 to 1.2 K day\(^{-1}\) if the heating were distributed over the 5 km deep boundary layer.

There is a significant correlation of 0.74 between TCWV and TOA net radiation. Figure 2a shows that TOA net downward radiation increases with TCWV (and associated dust and cloud), with a regression coefficient of +2.2 W kg\(^{-1}\). This is a result of a 3.2 W kg\(^{-1}\) increase in TOA net longwave with TCWV in observations (Fig. S1a in the Supplement), from water vapour, clouds (and associated dust) reducing TOA outgoing longwave. This longwave TCWV effect dominates the decrease in net shortwave with increased water vapour (−0.98 W kg\(^{-1}\), Fig. S1d), due to water vapour and associated clouds and dust. The correlations are strongest between TCWV and TOA net or longwave radiation (both 0.74 and 0.68), rather than TOA shortwave (−0.36), since the water vapour directly affects the longwave, while much
of the shortwave effects of TCWV are indirect, occurring via associated clouds and dust.

The correlation between AOD and TOA net radiation (Fig. 2b) is much weaker than between TCWV and TOA net radiation (0.26 compared with 0.74). Figure 2b shows that TOA net radiation increases with AOD (5.3 W m\(^{-2}\) per AOD, comparable with Balkanski et al., 2007), but this relationship is complex and its magnitude decreases to 3.5 W m\(^{-2}\) if all available data are used (with a correlation of 0.17). The increase in net TOA radiation with AOD occurs because the increase in TOA longwave (+10.5 W m\(^{-2}\) per AOD) dominates the decrease in TOA net shortwave (−5.2 W m\(^{-2}\) per AOD; Fig. S1b and e). The observed net effect of dust at TOA and the dominance of the longwave for this effect are both consistent with previous studies (Balkanski et al., 2007; Yang et al., 2009). Banks et al. (2014) show that in clear sky the diurnal mean effect of dust at BBM is warming in the shortwave. Therefore the observed reduced shortwave heating associated with dust reported in Fig. 2 is likely a result of cross correlation of AOD and cloud. This cloud, as well as the water vapour and dust, reduces outgoing longwave, leading to a warming. The effects of AOD and TCWV variations on radiation normalised by the standard deviation (\(\sigma\)) in either AOD or TCWV (Table 1, values in square brackets) show that the variance in TCWV has a much larger effect on TOA net radiation (10.4 W m\(^{-2}\) per \(\sigma\)) than the variance in AOD (3.6 W m\(^{-2}\) per \(\sigma\), or 2.3 W m\(^{-2}\) if “All data” are used), i.e. most day-to-day variations in net TOA radiation are mostly controlled by TCWV, not AOD.

Figure 2d shows daily net shortwave heating is always greater than net longwave cooling (the Earth-atmosphere system is warming in June). Daily variations in shortwave are anti-correlated with variations in longwave such that as daily net TOA shortwave decreases, the net longwave increases (correlation of −0.80). In Fig. 2d, if the gradient is less than −1, reducing the net shortwave will increase the net flux. The observed gradient is −1.4, i.e. days with net shortwave reduced by combinations of dust and cloud are associated with increased longwave heating (i.e. reduced longwave cooling) from the water vapour, dust and cloud that more than compensates for the decreased shortwave heating, resulting in greater net heating on these days. Figure 2d shows how there is greater variance in daily longwave cooling than shortwave warming and therefore, although they are coupled, variations in longwave cooling make the larger contribution to variations in TOA net radiation.

**TCWV and aerosol effects at TOA in ERA-I**

The ERA-I regression coefficients for TOA net radiation with TCWV of 1.3 W kg\(^{-1}\) (1.4 W kg\(^{-1}\) for all data) is similar to that observed (2.2 W kg\(^{-1}\), 2.1 W kg\(^{-1}\) for all data, Fig. 2c and a). ERA-I captures the sign of correlations of both TOA net longwave and shortwave with TCWV, although it un-
derestimates the magnitude of the regression coefficients for both (1.8 W kg\(^{-1}\) in longwave for ERA-I, compared with the 3.2 W kg\(^{-1}\) observed, and \(-0.48\) W kg\(^{-1}\) in shortwave for ERA-I compared with the \(-0.98\) W kg\(^{-1}\) observed; Fig. S1c and f). As observed, reduced net shortwave increases TOA net flux in ERA-I (Fig. 2e, gradient of \(-1.4\)).

Even though it does not account for the daily variations in dust, ERA-I captures much of the day-to-day variations in TOA net variation (correlations with observations are 0.62 and 0.73 for “All data” and “Good surface data”, not shown). Table 1 shows that the regression coefficients for ERA-I fluxes with observed AODs are of the correct sign: this suggests that some of the observed trends with AOD are due to associated water vapour and cloud (captured at least to some extent by ERA-I), rather than dust. This is consistent with the lower correlations between observed AOD and observed TOA net flux (0.26) than between observed TCWV and observed TOA net flux (0.74), discussed in the previous section.

The differences in the effects of TCWV in ERA-I and in observations are likely because of both errors in clouds in ERA-I and its lack of variability in dust. Detailed validation of model clouds over the bright dusty Sahara is challenging and beyond the scope of this paper. Here, we note that ERA captures day-to-day variations of mean cloud fraction (correlation with MPEF cloud mask of 0.56), but mean cloud fraction in ERA-I is 0.22, much less than the MPEF value of 0.53, although this value is likely biased high by dust. Surface albedo in ERA-I is very close to observed, but TOA upward shortwave in ERA-I is about 15 W m\(^{-2}\) less than in observations (although daily maxima in these values are similar, not shown). These comparisons with data both support the hypothesis that ERA-I underestimates cloud cover (consistent with Dolinar et al., 2015, Fig. 4). The underestimate of the regression coefficient of TOA net longwave with TCWV in ERA-I compared with observations (1.8 compared with 3.2 W kg\(^{-1}\)) is consistent with this suspected underestimation of cloud cover in ERA-I and also the lack of dust associated with TCWV reducing outgoing longwave (Haywood et al., 2005). However, in ERA-I the underestimate of the magnitude of the regression coefficient of TOA net longwave with TCWV (1.8 compared with 3.2 W kg\(^{-1}\)) and shortwave with TCWV (\(-0.48\) compared with \(-0.98\) W m\(^{-2}\)) compensate to some extent give a trend in TOA net radiation with TCWV of 1.3 W kg\(^{-1}\) in ERA-I, close to the 2.2 W kg\(^{-1}\) observed.

3.2 Control of surface net radiation by TCWV and AOD

At the surface there is a strong and significant decrease in net radiation with increasing AOD (Fig. 3b) with a regression coefficient of \(-13.1\) W m\(^{-2}\) per AOD. This is a result of compensating longwave and shortwave effects, with the shortwave effect being largest: Table 1 (and Fig. S2e) shows \(-31.9\) W m\(^{-2}\) surface net shortwave per AOD, with dust reducing solar heating at the surface (largely compensated by heating the atmosphere above, comparing with \(-5.2\) W m\(^{-2}\) TOA net shortwave per AOD, Sect. 3.3). Table 1 (Fig. S2b) shows \(+20.7\) W m\(^{-2}\) surface net longwave per AOD, i.e. dust, together with the water vapour and cloud associated with the dust, warms the surface in the longwave, but unlike at TOA this does not compensate fully for the shortwave effects. The effects of AOD on net, shortwave and longwave fluxes are consistent between the days with some interpolated values (asterisks) and other days (pluses).

TCWV decreases surface net radiation by 0.20 W kg\(^{-1}\) (Fig. 3a). This is a balance of \(+2.0\) W kg\(^{-1}\) from the longwave and \(-1.8\) W kg\(^{-1}\) from the shortwave, i.e. it is a small difference between two large numbers (Fig. S2a and d). Impacts of TCWV on surface net heating are therefore a subtle balance of water vapour, clouds and associated dust. If variations in surface net radiation with AOD and TCWV are normalised by the standard deviation in AOD or TCWV, variability in AOD is seen to dominate the variations in surface net radiation (square brackets in Table 1). For the impacts of TCWV, the days with some interpolated values at first appear to be inconsistent with other days (Figs. 3a, S1a, d), but this is due to the high AODs for these days, the effects of which are consistent with other data (Figs. Sb, S2be, e). At the surface, although the observed shortwave and longwave variations are anti-correlated (coefficient = \(-0.88\)), they cancel to a much lesser extent than at TOA. Figure 3d shows how decreased shortwave leads to increased net longwave, but this does not tend to compensate fully (gradient of \(-0.61\)), so decreased shortwave gives decreased net surface radiation. As such, daily variability in surface net radiation at BBM is influenced more by variability in the shortwave than the longwave. Again data from days with some interpolation of surface fluxes (squares) are consistent with other days (pluses).

The observed increase in surface net longwave with TCWV of 2.0 W kg\(^{-1}\) is within the range of 1.0 to 3.0 W kg\(^{-1}\) obtained by Evan et al. (2015) for Tamanrasset from observations, analyses and radiative transfer modelling. In summer at Tamanrasset TCWV might be expected to correlate with AOD as it does at BBM, and dust and clouds associated with TCWV in reality, but missing or under-estimated in analyses and radiative transfer modelling, may account for the greater sensitivity of surface net longwave to TCWV in observations compared with radiative transfer modelling and analyses, noted by Evan et al. (2015). The BBM value of 2.0 W kg\(^{-1}\) is slightly lower than the diurnal-mean observational value of 3.0 W kg\(^{-1}\) for Tamanrasset obtained by Evan et al. (2015), which may reflect the greater prevalence of clouds at the high-altitude Tamanrasset site, where mountains trigger moist convection (Birch et al., 2012). The BBM results also suggest that although the increases in net surface longwave with TCWV shown by Evan et al. (2015) could largely be compensated by coincident decreases in net surface shortwave (as at BBM), this is not expected at TOA.
supporting the proposed role by Evan et al. (2015) of water vapour in warming the SHL.

Effects of TCWV and AOD at the surface in ERA-I

Figure 3c shows that, in contrast with observations (Fig. 3a), ERA-I always produces an increase in net surface radiation with increasing TCWV (+0.76 W kg\(^{-1}\), compared with −0.20 W kg\(^{-1}\)). Figure 3e (gradient −1.1) shows that at the surface in ERA-I, unlike in observations, decreased net shortwave is always compensated by increased net longwave (i.e. reduced longwave cooling). This occurs since in ERA-I greater water vapour leads to greater net surface longwave (i.e. reduced longwave cooling, Fig. S2c), without the associated dust to reduce the net surface shortwave (Fig. S2f): the net surface radiation in ERA-I depends largely on surface longwave, whereas in observations it depends largely on the shortwave. As a result, ERA-I, which uses a monthly dust climatology, fails to capture day-to-day variations in surface net radiation, producing no correlation (0.02) with observations.

Although it does not affect the regression coefficients of surface fluxes with TCWV and AOD discussed above, we note here that ERA-I surface net longwave is on average 55 W m\(^{-2}\) less than observed, and this is almost all from more upward longwave than observed (not shown). Due to the non-linear nature of thermal emission, the 13 % error in upward longwave can be caused by only a 3 % error in skin temperature (or from an error in emissivity). Maximum values of daily ERA-I surface net shortwave are similar to observed, but minima are higher, likely from missing dust and cloud. These two errors lead to surface net radiation being around 34 W m\(^{-2}\) lower in ERA-I than observed.

3.3 Radiative heating of the atmosphere

The TOA and surface fluxes are differenced to give the radiative flux convergence within the atmosphere, i.e. the direct radiative heating of the atmosphere (Fig. 4). As expected the atmosphere is cooling in the longwave and is heated in the shortwave. There are statistically significant positive correlations between both TCWV or AOD (which are themselves correlated, Fig. 1) and net radiative heating of the atmosphere (Fig. 4a and b). This is consistent with the results of Slingo et al. (2006) and Slingo et al. (2009) for dust over the Sahel. For AOD there is a strong correlation (0.93) with shortwave atmospheric heating (Fig. S3e, 26.7 W m\(^{-2}\) per AOD, comparable with Balkanski et al., 2007) that dominates the trend of net longwave heating with AOD (Fig. S3c, −10.2 W m\(^{-2}\)). There are significant increases in net shortwave and net longwave radiative heating of the atmosphere with increasing TCWV (Fig. S3a and d, Table 1). The longwave effect (Fig. S3a) is much less clear than it is at TOA or at the surface, since the effects at TOA and the surface (Figs. S1a and S2a) are similar (3.2 and 2.0 W m\(^{-2}\)) and largely cancel each other.
When trends with TCWV and AOD are normalised by the standard deviations in TCWV and AOD to allow comparison (results in square brackets in Table 1), effects of AOD dominate those from TCWV, but this is much more pronounced in the shortwave. The results therefore show significant shortwave heating of the atmosphere by dust (consistent with Banks et al., 2014), consistent with the large effect of AOD on surface net and surface net shortwave fluxes, with much smaller effects at TOA. Decreases in surface heating associated with dust are largely compensated by direct radiative heating of the atmosphere. The shortwave heating from TCWV (correlation coefficient of only 0.19) is similar to that in ERA (below and Fig. S3f) showing that is not just from associated dust, but from shortwave absorption by water (although points with unusually high shortwave heating are explained by AODs, Fig. S3d and e). Figure 4d shows how increasing longwave cooling of the atmosphere is more than compensated for by the corresponding increased shortwave heating (gradient = \(-0.39\)); atmospheric heating is largely controlled by effects of dust on the shortwave, whereas longwave atmospheric heating is much less variable.

**Atmospheric heating in ERA-I**

ERA gives weaker longwave atmospheric cooling than observed and therefore less net atmospheric cooling (Figs. 4c and S3c). Lacking the observed variability in dust, ERA has little variability in atmospheric shortwave heating, with almost no correlation of shortwave heating with observed AODs (Table 1). ERA has a significant increase in shortwave atmospheric heating with TCWV (Fig. S3f, 0.91 W kg\(^{-1}\)) from absorption by water (similar to that observed, Fig. S3d, 0.78 W kg\(^{-1}\)). While observations have a significant, but weak, positive correlation between TCWV and longwave atmospheric heating (Fig. S3a, 0.34), ERA has a weak insignificant negative correlation (Fig. S3c, \(-0.20\)). Effects are weak in both cases, since TOA and surface longwave fluxes both respond similarly to TCWV and ERA is of course lacking the variability in dust that correlates with TCWV and this may contribute to the difference. Despite the weak variation in shortwave atmospheric heating in ERA compared with observations, variations in shortwave dominate the variations in net atmospheric heating, giving increased net heating with increased TCWV (Fig. 4c). This is however much weaker than observed (Fig. 4a), since ERA has much less variability in net heating due to its use of a dust climatology.

### 4 Discussion

Since variability in water dominates day-to-day variability in net TOA heating it is crucial for models to capture the water content of the SHL. Small errors in TCWV, in the altitude of the water vapour, or in associated cloud, could cause errors in clear-sky longwave radiation comparable with the 50 W m\(^{-2}\) from dust seen in Haywood et al. (2005), or the
20 to 40 W m$^{-2}$ model bias that Allan et al. (2011) show can be removed by the inclusion of dust. This paper shows that 50 W m$^{-2}$ TOA net longwave corresponds to around 16 kg m$^{-2}$ water (based on the 3.2 W kg$^{-1}$ dependence of TOA net longwave on TCWV, Table 1), roughly equivalent to 3 g kg$^{-1}$ over the 5 km deep boundary layer. Roberts et al. (2015) show that root mean square differences in analyses of WVMR at 20° N in the Sahara are around 1.5 g kg$^{-1}$, and show a case where differences between different analyses are around 4 g kg$^{-1}$. Garcia-Carreras et al. (2013) show a global model mean bias of around 1 g kg$^{-1}$ at Fennec supersite-1 in June 2011 in the model first guess (3-to-6 h forecast), despite assimilation of the Fennec radiosoundings. Models struggle to capture monsoon flow that cools and moistens the SHL, in particular from cold pools (Marsham et al., 2013b; Garcia-Carreras et al., 2013). This study shows that errors in fluxes of water vapour will lead to a compensating error of insufficient radiative heating from the absence of the moist air. Model errors in dust will affect the vertical distribution of heating and so also affect vertical mixing and dynamics.

The results give some insight into the Saharan BL energy budget during June over BBM. We show TOA net radiative heating of around 26 W m$^{-2}$. There was an observed mean night-time cooling of around 4 K over an approximately 1 km depth every night (Marsham et al., 2013a), corresponding to around 50 W m$^{-2}$ cooling (not all of this cooling is advective, some is radiative). To compensate for this cooling an additional warming of around 20 W m$^{-2}$ is required. Daily entrainment of free-tropospheric air will raise the BL top, which is lowered by subsidence to give, in the long term, a constant BL top. We can estimate the heating rate of the BL either from entrainment or subsidence. The 24 h entrainment flux is perhaps 10 W m$^{-2}$ (20 % of the 100 W m$^{-2}$ surface flux for 12 h). The 24 h subsidence of a lid of 5 K/100 m with 0.1 m s$^{-1}$ is 5 W m$^{-2}$. These simple estimates therefore leave a mis-match of around 10 W m$^{-2}$, but show that all terms (net daytime radiative warming, net night-time radiative and advective cooling, entrainment of warm subsiding air) are all of a similar order of magnitude and significant.

Although modelling is needed to fully understand the observed effects of water vapour on the radiation, the observations show that monsoon surges at BBM are expected to have significant effects on radiative heating rates. In June 2011 BBM experienced sudden moistenings of up to around 5 g kg$^{-1}$ (Fig. 5, Marsham et al., 2013a). If we assume that a value of 2.5 g kg$^{-1}$ is more representative of the change over the 5 km deep Saharan BL (Fig. 3 in Marshall et al., 2013a, shows such monsoon surges tend to directly affect the lower half of the 5 km layer) this gives a TOA net radiative heating of around 28 W m$^{-2}$ (based a TCWV of 12.5 kg m$^{-2}$ and a dependence of net radiation on TCWV of 2.2 W kg$^{-1}$, Table 1). If this heating is distributed over the 5 km deep Saharan BL it will result in a warming of around 0.5 K day$^{-1}$. It will therefore take days for the additional radiative warming to compensate for the cooling of a few degrees experienced in such events. This “radiative rewarming timescale” may be one contributing factor (together with timescales such as those for advection & mixing timescales and synoptic features such as African Easterly waves) to the variability of the 3-to-30-day variability of the SHL observed by Lavaysse et al. (2010).

The observed net radiative heating of the SHL region observed at BBM during June appears to contrast with Charney (1975), which shows heating from subsidence and TOA cooling from radiation for the Sahara in July. However, the Fennec supersite-1 is at the northern limit of the Inter Tropical Discontinuity and regularly receives cold moist air from the south (Marsham et al., 2013a). Charney (1975, Fig. 1) shows net TOA heating at the location of the Fennec supersite in July, with TOA net cooling only north of around 22° N (interestingly the TOA heating extends northeastwards over the Hoggar mountains, a region that favours northward extent of moist monsoon air; Cuesta et al., 2010). It is likely that further north away from the moistening from the monsoon the warmer drier atmosphere will give greater longwave cooling and a net radiative cooling, as shown by Charney (1975). This will be further investigated.

![Figure 5. Schematic showing net radiation and implied tropospheric radiative heating, in situations where either TCWV (top row) or AOD (bottom row) is perturbed by plus or minus one standard deviation away from their mean state (right and left columns respectively). Moist atmospheres tend to be dusty and vice versa. Red numbers show net shortwave, purple numbers show net longwave and black show net radiation. TOA and surface heating are shown by plus signs with downward arrows. Values are shown at surface, TOA and for inferred atmospheric radiative heating (“Atmos. Conv.”). Variance in TCWV has the dominant effect on net TOA radiation, while variance in AOD has the dominant effect on net surface radiation. Both TCWV and AOD are important for atmospheric heating rates.](image-url)
5 Conclusions

We have used unique observations of surface energy balance, TCWV and AOD from the central Sahara in June, together with retrievals from GERB, to investigate controls on the day-to-day variations in radiative heating in the SHL region. TOA fluxes show that on average the earth-atmosphere system is warming (26 W m$^{-2}$), the surface is warming (98 W m$^{-2}$) and the atmosphere is cooling (74 W m$^{-2}$), with the longwave cooling and the shortwave warming in each case. Although there are limits to the extent to which our empirical approach can disentangle the roles of dust, cloud and water vapour, largely due to correlations between these factors, the results provide new insight into their roles in controlling the radiative balance of the unique environment of the central Sahara (schematic in Fig. 5).

Water vapour and dust are observed to correlate in the central Sahara, likely due to the uplift of dust in monsoon surges and haboobs (Bou Karam et al., 2008; Marsham et al., 2008, 2013a). However, variations in water vapour (and associated variables such as temperature and cloud) and not variations in dust dominates day-to-day variability of TOA net radiation, and hence total heating of the earth-atmosphere system. ERA-I captures the observed variation in TOA net radiation (correlation with observations of around 0.65), despite a monthly dust climatology in ERA-I, which cannot capture day-to-day variations in dustiness. Variations in AOD dominate day-to-day variations in surface net radiation, which unsurprisingly are not captured in ERA-I. If effects from TCWV were simply due to correlated changes in AOD, or vice versa, these contrasting roles of TCWV and AOD at the TOA and surface would not be so distinct.

At TOA, on average, decreased shortwave heating gives greater net heating due to associated increases in longwave heating. ERA-I captures this and the overall impact of TCWV on TOA net radiation, with a mean increase in shortwave heating. ERA-I gives the correct TOA net radiation in response to TCWV and AOD at the TOA and surface would not be so distinct.

The results show that, when the SHL is cooled by cold moist air from its margins, the overall effect is to increase net TOA radiative heating, re warming the SHL, a feedback which stabilises the system, by re warming the cool air. This occurs in both reality and ERA-I. This ventilation by cold air is, however, normally accompanied by clouds and dust, which together reduce surface net radiation, which is not captured by ERA-I, as ERA-I is missing the variations in dust (and likely under-predicts cloudiness). As a result, even if ERA-I gives the correct TOA net radiation in response to water vapour, it fails to distribute this heating correctly in the vertical, with too much surface heating and insufficient boundary-layer heating. This will destabilise the boundary layer profile compared with reality, affecting subsequent modelled dry and moist convection and therefore modelled transport of heat, momentum, water vapour and dust.

Improved modelling of the energy budget of the SHL region is needed in models to improve predictions of the WAM across timescales (e.g. Evan et al., 2015). The results show that it is important that models used for predictions can accurately capture the processes controlling the water vapour distribution over the Sahara, as well as the dust. This capability is currently questionable for water (Marsham et al., 2013b; Birch et al., 2014; Garcia-Carreras et al., 2013; Roberts et al., 2015), clouds (Roehrig et al., 2013; Stein et al., 2015) and dust (Evan et al., 2014), with many dust errors coming from moist convection (Marsham et al., 2011; Heinold et al., 2013). The results presented here therefore strongly motivate the need to improve the representation of advection of water vapour, clouds and convection in models.

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