

1 The contrasting roles of water and dust in controlling daily variations
2 in radiative heating of the summertime Saharan Heat Low

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12
13 **Abstract**

14 The summertime Sahara Heat Low (SHL) is a key component of the West African Monsoon (WAM)
15 system. Considerable uncertainty remains over the relative roles of water vapour and dust aerosols in
16 controlling the radiation budget over the Sahara and therefore our ability to explain variability and
17 trends in the SHL, and in turn, the WAM. Here, new observations from Fennec supersite-1 in the
18 central Sahara during June 2011 and June 2012, together with satellite retrievals from GERB, are
19 used to quantify how total column water vapour (TCWV) and dust aerosols (from aerosol optical
20 depth, AOD) control day-to-day variations in energy balance in both observations and ECWMF
21 reanalyses (ERA-I). The data show that the earth-atmosphere system is radiatively heated in June
22 2011 and 2012. Although the empirical analysis of observational data cannot completely disentangle
23 the roles of water vapour, clouds and dust, the analysis demonstrates that TCWV provides a far
24 stronger control on TOA net radiation, and so the net heating of the earth-atmosphere system, than
25 AOD does. In contrast, variations in dust provide a much stronger control on surface heating, but the
26 decreased surface heating associated with dust is largely compensated by increased atmospheric
27 heating, and so dust control on net TOA radiation is weak. Dust and TCWV are both important for
28 direct atmospheric heating. ERA-I, which assimilated radiosondes from the Fennec campaign,
29 captures the control of TOA net flux by TCWV, with a positive correlation ($r=0.6$) between observed
30 and modelled TOA net radiation, despite the use of a monthly dust climatology in ERA-I that cannot

1 capture the daily variations in dustiness. Variations in surface net radiation, and so the vertical profile
2 of radiative heating, are not captured in ERA-I, since it does not capture variations in dust. Results
3 show that ventilation of the SHL by cool moist air leads to a radiative warming, stabilising the SHL
4 with respect to such perturbations. It is known that models struggle to capture the advective
5 moistening of the SHL, especially that associated with mesoscale convective systems. Our results
6 show that the typical model errors in Saharan water vapour will lead to substantial errors in the
7 modelled TOA energy balance (tens of W m^{-2}), which will lead to errors in both the SHL and the
8 WAM.

9 **1 Introduction**

10 The Sahara lies under the descending branch of the Hadley circulation and during summer the intense
11 solar heating combined with the arid environment leads to large sensible surface heat fluxes and the
12 formation the Saharan Heat Low (SHL). This increases the pressure gradient from the Gulf of Guinea
13 to the Sahara, driving the West African Monsoon (WAM), and variations in the SHL modify the
14 WAM on time scales from days to decades [Thorncroft and Blackburn 1999; Peyrillé and Lafore
15 2007; Biasutti et al., 2009, Lavaysse et al, 2009, 2010; Chauvin et al., 2010, Xue et al., 2010, Martin
16 and Thorncroft 2014, Martin et al., 2014, Dong and Sutton, 2015]. There is a shortage of routine
17 observations in the SHL and substantial disagreements exist even between analyses [Marsham et al.,
18 2011; Roberts et al., 2014]. The Fennec project aimed to better quantify processes governing the
19 Saharan atmosphere [Washington et al., 2012; Ryder et al., 2015] and deployed an observational
20 supersite-1 close to the climatological centre of the SHL [Marsham et al., 2013a].

21 The radiative budget of the Sahara is significantly modulated by variations in clouds, dust and water
22 vapour. Charney (1975) shows how the high albedo and dry atmosphere can lead to a top-of-
23 atmosphere (TOA) net radiative cooling in July, with heating via subsidence, proposing a positive
24 feedback where dry soils with little vegetation generate high albedo, favouring atmospheric descent
25 and low rainfall. The dry atmosphere means that water vapour provides a key control in the
26 longwave with vapour at all levels affecting top-of-atmosphere (TOA) outgoing longwave (Allan et
27 al., 1999; Brindley and Harries, 1998). Evan et al. (2015) suggest that the increasing temperatures
28 within the SHL over the past 30 years, key to the recovery of the Sahel from drought, are driven by
29 longwave impacts of increasing water vapour, in the “Saharan Water Temperature” feedback and
30 Dong and Sutton (2015) propose a greenhouse-gas driven increase with a feedback through water
31 vapour. Shallow clouds on top of the deep dry boundary layer [Cuesta et al., 2009] occur around
32 20% of the time over the Sahara, with mid-level clouds reducing net surface shortwave and
33 increasing net surface longwave in the Sahel [Stein et al., 2011; Bouniol et al., 2011]. Dust absorbs
34 and emits longwave radiation (Haywood et al., 2005) and scatters and absorbs shortwave [Ryder et

1 al., 2013; Banks et al., 2014]. At the TOA, and over the bright Sahara, dust induces a warming as its
2 longwave effects dominate its shortwave effects (Balkanski et al., 2007; Yang et al., 2009).
3 Operational models use either prognostic dust or dust climatologies, but struggle to capture
4 variations in summertime dust, partly as cold-pool outflows from convection (haboobs) provide a
5 key uplift mechanism that is missing in operational models that use parametrised moist convection
6 [Marsham et al, 2011; Heinold et al., 2013; Marsham et al., 2013a].

7 At low levels the Sahara is cooled by advection from neighbouring moister and cooler regions,
8 including the WAM to the south. Representing the monsoon is a challenge to models, partly because
9 of the representation of convection, in particular its diurnal timing and cold pools; the diurnal timing
10 of Sahelian moist convection affects the pressure gradient driving the monsoon, modulating the flux
11 of water vapour from the Sahel to the Sahara and hence rainfall over the Sahel (Marsham et al., 2013b;
12 Birch et al., 2014). Furthermore cold pools form a significant component of the monsoon (Marsham
13 et al., 2013b) and also ventilate the Sahara from the Atlas in the north [Emmel et al., 2010]; most
14 temperature and humidity biases in the Met Office global model at the Fennec supersite-1 during June
15 2011 were caused by missing cold pool advection [Garcia-Carreras et al., 2013]. Similarly ventilation
16 of the Sahara by the Atlantic Inflow involves mesoscale flows that are a challenge for global models
17 [Grams et al., 2010; Todd et al., 2013]. Since clouds, water vapour and dust are all important to the
18 Sahara's radiative energy balance, such model errors in convection, clouds, haboobs and advection
19 of water vapour will all affect modelled radiative energy balances and hence climate.

20 There is a clear need to establish the controls on the radiation budget over the Sahara and evaluate
21 models. In this paper we use observations of surface radiative fluxes from Fennec supersite-1 in the
22 central Sahara and retrievals of TOA fluxes from satellite data to investigate how dust and water
23 together control the day-to-day variations in energy balance over the Fennec supersite-1 in the
24 summertime SHL region, and how this is represented in ERA-Interim (ERA-I) reanalysis. Results in
25 Section 3 show that TCWV and AOD are correlated and we cannot completely isolate the effects of
26 either TCWV or dust. However, TCWV and AOD have sufficiently independent variations, and
27 sufficiently distinct impacts at solar and infrared wavelengths, which conform with physical
28 principles, that the results give unique insights into their contrasting roles in the central Sahara.
29 Section 2 describes methods, section 3 presents results, section 4 contains discussion and conclusions
30 are in section 5.

31 **2 Method**

32 We use data from Fennec supersite-1 in the central Sahara, located at Bordj-Badji Mokhtar (BBM) at
33 21.4N 0.9E (in the very south of Algeria, close to the triple point of Algeria, Mali and Niger), close

1 to the SHL's climatological centre and the dust maximum [Marsham et al., 2013a], together with
2 corresponding values from ERA-Interim [Dee et al., 2011], and satellite TOA fluxes (Harries et al.,
3 2005; Dewitte et al., 2008). These satellite TOA fluxes are produced using a narrow to broad band
4 conversion of SEVIRI (Spinning Enhanced Visible and Infrared Imager) radiance measurements.
5 These are scaled by co-located GERB (Geostationary Earth Radiation Budget experiment)
6 measurements and converted into broadband fluxes measurements at a horizontal resolution of 3×3
7 SEVIRI pixels (0.32-4 microns in the shortwave, and 4-100 microns in the longwave). This
8 enhancement gives a spatial resolution of 9km at nadir, compared with the 45km of the native GERB.
9 These observed fluxes are from clear and cloudy skies, but we also use the European Organisation
10 for the Exploitation of Meteorological Satellites' MPEF (Meteorological Product Extraction Facility)
11 cloud mask as a simple measure of cloud cover. ERA-Interim uses aerosol climatologies so cannot
12 capture day-to-day variations in dust. Radiosonde data from the Fennec supersite were assimilated
13 into ERA-I, which will have improved its representation of the thermodynamic profile (see Garcia-
14 Carreras et al., 2015 for impacts of assimilation of Fennec radiosondes on the Met Office global
15 forecast model).

16 Fennec data are from intensive observation periods (IOPs) in June 2011 and 2012, when a Cimel sun
17 photometer provided Aerosol Optical Depths (AODs) at 675 nm, a Kipp & Zonen radiometer
18 mounted at 2 m provided measurements of broad-band radiative fluxes and 3 to 6-hourly radiosonde
19 observations were available (Marsham et al., 2013a). The sun-photometer is part of the AERONET
20 program (Holben et al., 1998) and cloud-screened AOD retrievals are only available during the day.
21 Level-2 AOD data are not available for 2012 since not all data meet level-2 requirements. However,
22 the 0.675 nm AODs are still reliable. We therefore use level-2 data for 2011 and level-1.5 for 2012,
23 noting that using only 2011 data does not affect our conclusions. We use the radiosondes to compute
24 column water vapour from the surface to 300 hPa (a height consistently reached by the radiosondes),
25 which we refer to simply as "total column water vapour (TCWV)". During June 2011 BBM was
26 regularly cooled by nocturnal monsoon flows and embedded cold pools giving substantial variability
27 in TCWV [Marsham et al., 2013a], and qualitatively similar weather events were observed during
28 June 2012.

29 In order to study the day-to-day variations in the energy budget, we average all data to their daily
30 means. Complete surface flux data were only available for 11 days in June 2011 (9, 10, 18-20, 23-27,
31 30 June) and 25 days in June 2012 (all except 4, 16-18, 30 June). Some dates had short data gaps
32 (around two hours on two days, but otherwise an hour or less) and these gaps were interpolated across
33 in order to include 7, 17, 21, 22 June 2011 and 16-17 June 2012. This gave an improved range of
34 AODs, albeit with increased uncertainties in surface fluxes. Fluxes from these days with some

1 interpolation are marked by squares in Figures 1 to 4, and the effects of interpolation are discussed in
2 Section 3, where it is seen that results from these days are physically consistent with other data from
3 days without interpolation. The surface flux data from Kipp and Zonen radiometers have slightly
4 different spectral ranges to the satellite-borne GERB: Kipp and Zonen are 0.3 to 2.8 μm in the
5 shortwave and 4.5 to 42 μm in the longwave, whilst GERB is 0.32 to 4 μm and 4 to 100 μm . This
6 means that the surface-based Kipp and Zonen can miss up to 3.5 W m^{-2} net shortwave atmospheric
7 heating as would be seen by GERB and up to 3.8 W m^{-2} of the net longwave as would be seen by
8 GERB (Banks et al., 2014). This introduces errors of up to 4 W m^{-2} in our inferred direct atmospheric
9 radiative heating rates, but does not affect our analysis and conclusions, which is focused on the
10 controls on the variability of these rates, rather than their absolute values.

11 Sun-photometer AODs were available from 8 June 2011 (with no observations on 13 June) and 1 to
12 28 June 2012 (with no observations on 17-19 June). Radiosondes were available from 8 June 2011
13 and 1 to 26 June 2012. The number of observations contributing to the daily mean is variable for
14 AODs, since observations are only made when it is cloud free, but all days except one had at least
15 eight AOD observations and the daily-mean AOD range of 0.2 to 2.7 is similar to that of the
16 observation range in AODs (0.2 to 3.9) and the diurnal cycle in AOD is weak (Marsham et al., 2013a;
17 Banks et al., 2014). Overall this gave 36 days with surface data, observed AOD and observed TCWV
18 and 44 days with TCWV and AOD.

19 **3 Results**

20 In order to determine how the changing amounts of water and dust over BBM affect the changing
21 radiative heating at the surface, TOA and within the atmosphere we analyse relationships between
22 the daily means of key variables, using both observed quantities and the equivalent from ERA-I at
23 the location of BBM. ERA-I uses a climatological AOD field and so cannot capture the observed
24 daily variability in AOD. This, in effect, represents a quasi-control experiment for dust variability.
25 Net fluxes are defined as downward, with increased net downward flux corresponding to increased
26 shortwave heating or reduced longwave cooling. All correlations and slopes of linear regression lines
27 discussed are listed in Table 1 (correlations in bold are significant at the 90% level). Relationships
28 are shown using days where surface data are available (referred to as “Good surface data”), and for
29 all available data (“All data”) where surface flux data are not required. The regressions are very
30 similar whichever dataset is used and values in the text are for “Good surface data”, unless otherwise
31 noted. The effect of subsampling is small for ERA-I, showing that general lessons can be drawn from
32 the observational data, despite the limited time-span of the dataset. Similarly, for both June 2011 and
33 2012 analysed water vapour at 850 and 925 hPa and AODs from MISR, Deep blue (Terra and Aqua)

1 and OMI are all within one standard deviation of their mean values (not shown) and there is no
2 indication that the weather regimes affecting BBM in these periods were anomalous.

3 There are correlations between dust and water (discussed below) which mean that effects of either
4 cannot be completely isolated from the other, but nevertheless the approach allows identification of
5 how variations in these variables affect radiative heating. Figure 1 shows that there is a significant
6 tendency for more dust with more TCWV, although there are a few dry dusty days (correlation =
7 0.29). The use of surface flux data with some interpolation (shown by squares) allows study of more
8 days with high AODs. The mechanisms underlying this correlation are understood: Marsham et al.
9 (2013a) shows how moist monsoon surges from the south are associated with dust at BBM. This is
10 because the moist surges are associated with both dusty haboobs and moist nocturnal low-level jets
11 (LLJs) that together dominate the dust uplift at BBM in June 2011 [Marsham et al., 2013a; Allen et
12 al., 2013]. The association between dust and water vapour is consistent with Figure 16 in Marsham
13 et al. (2013a), which shows a statistical link between AOD and cloud cover at BBM. Intense dust
14 uplift does sometimes occur in dry air, however, mainly in the dry Harmattan LLJs [Marsham et al.,
15 2013a; Allen et al., 2013].

16 **3.1 Control of TOA net radiation by water (TCWV) and aerosols (AOD)**

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

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

31 The correlation between AOD and TOA net radiation (Figure 2b) is much weaker than between
32 TCWV and TOA net radiation (0.26 compared with 0.74). Figure 2b shows that TOA net radiation
33 increases with AOD (5.3 W m^{-2} per AOD, comparable with Balkanski et al., 2007), but this

1 relationship is complex and its magnitude decreases to 3.5 W m^{-2} if all available data are used (with a
2 correlation of 0.17). The increase in net TOA radiation with AOD occurs because the increase in
3 TOA longwave ($+10.5 \text{ W m}^{-2}$ per AOD) dominates the decrease in TOA net shortwave (-5.2 W m^{-2}
4 per AOD; Figures S1b and S1e). The observed net effect of dust at TOA and the dominance of the
5 longwave for this effect are both consistent with previous studies (Balkanski et al. 2007; Yang et al.,
6 2009). Banks et al. (2014) show that in clear-sky the diurnal mean effect of dust at BBM is warming
7 in the shortwave. Therefore the observed reduced shortwave heating associated with dust reported in
8 Figure 2 is likely a result of cross correlation of AOD and cloud. This cloud, as well as the water
9 vapour and dust, reduces outgoing longwave, leading to a warming. The effects of AOD and TCWV
10 variations on radiation normalised by the standard deviation (σ) in either AOD or TCWV (Table 1,
11 values in square brackets) show that the variance in TCWV has a much larger effect on TOA net
12 radiation (10.4 W m^{-2} per σ) than the variance in AOD (3.6 W m^{-2} per σ , or 2.3 W m^{-2} if “All data”
13 are used), *i.e.* most day-to-day variations in net TOA radiation are mostly controlled by TCWV, not
14 AOD.

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

26 **3.1.1 TCWV and aerosol effects at TOA in ERA-I**

27 The ERA-I regression coefficients for TOA net radiation with TCWV of 1.3 W kg^{-1} (1.4 W kg^{-1} for
28 all data) is similar to that observed (2.2 W kg^{-1} , 2.1 W kg^{-1} for all data, Figures 2c and 2a). ERA-I
29 captures the sign of correlations of both TOA net longwave and shortwave with TCWV, although it
30 underestimates the magnitude of the regression coefficients for both (1.8 W kg^{-1} in longwave for
31 ERA-I, compared with the 3.2 W kg^{-1} observed, and -0.48 W kg^{-1} in shortwave for ERA-I compared
32 with the -0.98 W kg^{-1} observed; Figures S1c and S1f). As observed, reduced net shortwave increases
33 TOA net flux in ERA-I (Figure 2e, gradient of -1.4).

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

9 The differences in the effects of TCWV in ERA-I and in observations are likely because of both errors
10 in clouds in ERA-I and its lack of variability in dust. Detailed validation of model clouds over the
11 bright dusty Sahara is challenging and beyond the scope of this paper. Here, we note that ERA
12 captures day-to-day variations of mean cloud fraction (correlation with MPEF cloud mask of 0.56),
13 but mean cloud fraction in ERA-I is 0.22, much less than the MPEF value of 0.53, although this
14 value is likely biased high by dust. Surface albedo in ERA-I is very close to observed, but TOA
15 upward shortwave in ERA-I is about 15 W m^{-2} less than in observations (although daily maxima in
16 these values are similar, not shown). These comparisons with data both support the hypothesis that
17 ERA-I underestimates cloud cover (consistent with Dolinar et al. (2015) Figure 4). The underestimate
18 of the regression coefficient of TOA net longwave with TCWV in ERA-I compared with observations
19 (1.8 compared with 3.2 W kg^{-1}) is consistent with this suspected underestimation of cloud cover in
20 ERA-I and also the lack of dust associated with TCWV reducing outgoing longwave (Haywood et
21 al., 2005). However, in ERA-I the underestimation of the magnitude of the regression coefficient of
22 TOA net longwave with TCWV (1.8 compared with 3.2 W kg^{-1}) and shortwave with TCWV (-0.48
23 compared with -0.98 W m^{-2}) compensate to some extent give a trend in TOA net radiation with TCWV
24 of 1.3 W kg^{-1} in ERA-I, close to the 2.2 W kg^{-1} observed.

25 **3.2 Control of surface net radiation by TCWV and AOD**

26 At the surface there is a strong and significant decrease in net radiation with increasing AOD (Figure
27 3b) with a regression coefficient of -13.1 W m^{-2} per AOD. This is a result of compensating longwave
28 and shortwave effects, with the shortwave effect being largest: Table 1 (and Figure S2e) shows -31.9
29 W m^{-2} surface net shortwave per AOD, with dust reducing solar heating at the surface (largely
30 compensated by heating the atmosphere above, comparing with -5.2 W m^{-2} TOA net shortwave per
31 AOD, Section 3.3). Table 1 (Figure S2b) shows $+20.7 \text{ W m}^{-2}$ surface net longwave per AOD, *i.e.*
32 dust, together with the water vapour and cloud associated with the dust, warms the surface in the
33 longwave, but unlike at TOA this does not compensate fully for the shortwave effects. The effects of

1 AOD on net, shortwave and longwave fluxes are consistent between the days with some interpolated
2 values (asterisks) and other days (pluses).

3 TCWV decreases surface net radiation by 0.20 W kg^{-1} (Figure 3a). This is a balance of $+2.0 \text{ W kg}^{-1}$
4 from the longwave and -1.8 W kg^{-1} from the shortwave i.e. is a small difference between two large
5 numbers (Figures S2a and S2d). Impacts of TCWV on surface net heating are therefore a subtle
6 balance of water vapour, clouds and associated dust. If variations in surface net radiation with AOD
7 and TCWV are normalised by the standard deviation in AOD or TCWV, variability in AOD is seen
8 to dominate the variations in surface net radiation (square brackets in Table 1). For the impacts of
9 TCWV, the days with some interpolated values at first appear to be inconsistent with other days
10 (Figures 3a, S1a, S1d), but this is due to the high AODs for these days, the effects of which are
11 consistent with other data (Figures 3b, S2be, S2e). At the surface, although the observed shortwave
12 and longwave variations are anti-correlated (coefficient = -0.88), they cancel to a much lesser extent
13 than at TOA. Figure 3d shows how decreased shortwave leads to increased net longwave, but this
14 does not tend to compensate fully (gradient of -0.61), so decreased shortwaves gives decreased net
15 surface radiation. As such, daily variability in surface net radiation at BBM is influenced more by
16 variability in the shortwave than the longwave. Again data from days with some interpolation of
17 surface fluxes (squares) are consistent with other days (pluses).

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

31 **3.2.1 Effects of TCWV and AOD at the surface in ERA-I**

32 Figure 3c shows that, in contrast with observations (Figure 3a), ERA-I always produces an *increase*
33 in net surface radiation with increasing TCWV ($+0.76 \text{ W kg}^{-1}$, compared with -0.20 W kg^{-1}). Figure

1 3e (gradient -1.1) shows that at the surface in ERA-I, unlike in observations, decreased net shortwave
2 is always compensated by increased net longwave (i.e. reduced longwave cooling). This occurs since
3 in ERA-I greater water vapour leads to greater net surface longwave (i.e. reduced longwave cooling,
4 Figure S2c), without the associated dust to reduce the net surface shortwave (Figure S2f): the net
5 surface radiation in ERA-I depends largely on surface longwave, whereas in observations it depends
6 largely on the shortwave. As a result, ERA-I, which uses a monthly dust climatology, fails to capture
7 day-to-day variations in surface net radiation, producing no correlation (0.02) with observations.

8 Although it does not affect the regression coefficients of surface fluxes with TCWV and AOD
9 discussed above, we note here that ERA-I surface net longwave is on average 55 W m^{-2} less than
10 observed, and this is almost all from more upward longwave than observed (not shown). Due to the
11 non-linear nature of thermal emission, the 13% error in upward longwave can be caused by only a
12 3% error in skin temperature (or from an error in emissivity). Maximum values of daily ERA-I surface
13 net shortwave are similar to observed, but minima are higher, likely from missing dust and cloud.
14 These two errors lead to surface net radiation being around 34 W m^{-2} lower in ERA-I than observed.

15 **3.3 Radiative heating of the atmosphere**

16 The TOA and surface fluxes are differenced to give the radiative flux convergence within the
17 atmosphere, *i.e.* the direct radiative heating of the atmosphere (Figure 4). As expected the atmosphere
18 is cooling in the longwave and is heated in the shortwave. There are statistically significant positive
19 correlations between both TCWV or AOD (which are themselves correlated, Figure 1) and net
20 radiative heating of the atmosphere (Figures 4a and 4b). This is consistent with the results of Slingo
21 et al. (2006) and Slingo et al., (2009) for dust over the Sahel. For AOD there is a strong correlation
22 (0.93) with shortwave atmospheric heating (Figure S3e, 26.7 W m^{-2} per AOD, comparable with
23 Balkanski et al., 2007) that dominates the trend of net longwave heating with AOD (Figure S3c, -
24 10.2 W m^{-2}). There are significant increases in net shortwave and net longwave radiative heating of
25 the atmosphere with increasing TCWV (Figures S3a and S3d, Table 1). The longwave effect (Fig.
26 S3a) is much less clear than it is at TOA or at the surface, since the effects at TOA and the surface
27 (Figures S1a and S2a) are similar (3.2 and 2.0 W m^{-2}) and largely cancel.

28 When trends with TCWV and AOD are normalised by the standard deviations in TCWV and AOD
29 to allow comparison (results in square brackets in Table 1), effects of AOD dominate those from
30 TCWV, but this is much more pronounced in the shortwave. The results therefore show significant
31 shortwave heating of the atmosphere by dust (consistent with Banks et al., 2014), consistent with the
32 large effect of AOD on surface net and surface net shortwave fluxes, with much smaller effects at
33 TOA. Decreases in surface heating associated with dust are largely compensated by direct radiative

1 heating of the atmosphere. The shortwave heating from TCWV (correlation coefficient of only 0.19)
2 is similar to that in ERA (below and Fig. S3f) showing that is not just from associated dust, but from
3 shortwave absorption by water (although points with unusually high shortwave heating are explained
4 by AODs, Figures S3d and S3e). Figure 4d shows how increasing longwave cooling of the
5 atmosphere is more than compensated for by the corresponding increased shortwave heating (gradient
6 = -0.39); atmospheric heating is largely controlled by effects of dust on the shortwave, whereas
7 longwave atmospheric heating is much less variable.

8 **3.3.1 Atmospheric heating in ERA-I**

9 ERA gives weaker longwave atmospheric cooling than observed and therefore less net atmospheric
10 cooling (Figures 4c and S3c). Lacking the observed variability in dust, ERA has little variability in
11 atmospheric shortwave heating, with almost no correlation of shortwave heating with observed AODs
12 (Table 1). ERA has a significant increase in shortwave atmospheric heating with TCWV (Figure S3f,
13 0.91 W kg^{-1}) from absorption by water (similar to that observed, Figure S3d, 0.78 W kg^{-1}). While
14 observations have a significant, but weak, positive correlation between TCWV and longwave
15 atmospheric heating (Figure S3a, 0.34), ERA has a weak insignificant negative correlation (Figure
16 S3c, -0.20). Effects are weak in both cases, since TOA and surface longwave fluxes both respond
17 similarly to TCWV and ERA is of course lacking the variability in dust that correlates with TCWV
18 and this may contribute to the difference. Despite the weak variation in shortwave atmospheric
19 heating in ERA compared with observations, variations in shortwave dominate the variations in net
20 atmospheric heating, giving increased net heating with increased TCWV (Figure 4c). This is however
21 much weaker than observed (Figure 4a), since ERA has much less variability in net heating due to its
22 use of a dust climatology.

23 **4 Discussion**

24 Since variability in water dominates day-to-day variability in net TOA heating it is crucial for models
25 to capture the water content of the SHL. Small errors in TCWV, in the altitude of the water vapour,
26 or in associated cloud, could cause errors in clear-sky longwave radiation comparable with the 50 W
27 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)
28 show can be removed by the inclusion of dust. This paper shows that 50 W m^{-2} TOA net longwave
29 corresponds to around 16 kg m^{-2} water (based on the 3.2 W kg^{-1} dependence of TOA net longwave
30 on TCWV, Table 1), roughly equivalent to 3 g kg^{-1} over the 5-km deep boundary layer. Roberts et al.
31 (2014) show that route-mean-square differences in *analyses* of WVMR at 20N in the Sahara are
32 around 1.5 g kg^{-1} , and show a case where differences between different analyses are around 4 g kg^{-1} .
33 Garcia-Carreras et al. (2013) show a global model mean bias of around 1 g kg^{-1} at Fennec supersite-

1 1 in June 2011 in the model first guess (3-to-6 hour forecast), despite assimilation of the Fennec
2 radiosoundings. Models struggle to capture monsoon flow that cools and moistens the SHL, in
3 particular from cold pools [Marsham et al., 2013b; Garcia-Carreras et al., 2013]. This study shows
4 that errors in fluxes of water vapour will lead to a compensating error of insufficient radiative heating
5 from the absence of the moister air. Model errors in dust will affect the vertical distribution of heating
6 and so also affect vertical mixing and dynamics.

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

19
20 Although modelling is needed to fully understand the observed effects of water vapour on the
21 radiation, the observations show that monsoon surges at BBM are expected to have significant effects
22 on radiative heating rates. In June 2011 BBM experienced sudden moistenings of up to around 5 g
23 kg^{-1} (Fig. 5, Marsham et al., 2013a). If we assume that a value of 2.5 g kg^{-1} is more representative of
24 the change over the 5 km deep Saharan BL (Fig. 3 in Marsham et al. (2013a) shows such monsoon
25 surges tend to directly affect the lower of half of the 5-km layer) this gives a TOA net radiative heating
26 of around 28 W m^{-2} (based a TCWV of 12.5 g m^{-2} and a dependence of net radiation on TCWV of
27 2.2 W kg^{-1} , Table 1). If this heating is distributed over the 5 km deep Saharan BL it will result in a
28 warming of around 0.5 K day^{-1} . It will therefore take days for the additional radiative warming to
29 compensate for the cooling of a few degrees experienced in such events. This “radiative rewarming
30 time scale” may be one contributing factor (together with time-scales such as those for advection &
31 mixing time scales and synoptic features such as African Easterly waves) to the variability of the 3-
32 to-30-day variability of the SHL observed by Lavaysse et al (2010).

33 The observed net radiative heating of the SHL region observed at BBM during June appears to
34 contrast with Charney (1975), which shows heating from subsidence and TOA cooling from radiation

1 for the Sahara in July. However, the Fennec supersite-1 is at the northern limit of the Inter Tropical
2 Discontinuity and regularly receives cold moist air from the south [Marsham et al., 2013a]. Charney
3 (1975) Figure 1 shows net TOA heating at the location of the Fennec supersite in July, with TOA net
4 cooling only north of around 22N (interestingly the TOA heating extends northeastwards over the
5 Hoggar mountains, a region that favours northward extent of moist monsoon air [Cuesta et al., 2010]).
6 It is likely that further north away from the moistening from the monsoon the warmer drier
7 atmosphere will give greater longwave cooling and a net radiative cooling, as shown by Charney
8 (1975). This will be further investigated.

9

10 **5 Conclusions**

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

20 Water vapour and dust are observed to correlate in the central Sahara, likely due to the uplift of dust
21 in monsoon surges and haboobs (Bou Karam et al., 2008; Marsham et al., 2008; Marsham et al.,
22 2013a). However, variations in water vapour (and associated variables such as temperature and cloud)
23 and not variations in dust dominates day-to-day variability of TOA net radiation, and hence total
24 heating of the earth-atmosphere system. ERA-I captures the observed variation in TOA net radiation
25 (correlation with observations of around 0.65), despite a monthly dust climatology in ERA-I, which
26 cannot capture day-to-day variations in dustiness. Variations in AOD dominate day-to-day variations
27 in surface net radiation, which unsurprisingly are not captured in ERA-I. If effects from TCWV were
28 simply due to correlated changes in AOD, or visa versa, these contrasting roles of TCWV and AOD
29 at the TOA and surface would not be so distinct.

30 At TOA, on average, decreased shortwave heating gives greater net heating due to associated
31 increases in longwave heating. ERA-I captures this and the overall impact of TCWV on TOA net
32 radiation, with a mean increase in TOA net radiation with TCWV of 1.3 W kg^{-1} compared with 2.2
33 W kg^{-1} in observations. There are, however, compensating errors in the effects of TOA net shortwave

1 and longwave with TCWV in ERA-I. ERA-I under-estimates the effects of TCWV on both TOA
2 longwave and shortwave: it misses corresponding variations in dust and although it captures much of
3 the effects of water vapour, it likely underestimates cloud (and significant uncertainties in analysed
4 water vapour persist at BBM, even when radiosondes are assimilated, Garcia-Carreras et al., 2013).

5 At the surface, dust (and associated water vapour and cloud) decreases net surface radiation in reality
6 by around 13 W m^{-2} per AOD. Although increasing TCWV reduces the surface longwave cooling,
7 the effect of TCWV on the net surface radiation is weak, variable and a subtle balance between the
8 competing effects of water vapour, clouds and dust (-0.2 W kg^{-1}). Unlike at the TOA, at the surface
9 decreases in shortwave are on average not compensated by increases in longwave, leading to
10 decreased net radiation with decreased shortwave. In contrast to the observations, ERA-I gives greater
11 net surface radiation with decreased surface shortwave: it is missing the effects of varying dust and
12 can only capture the effects of water and cloud, likely underestimating cloud. This gives no
13 correlation between ERA-I surface net radiation and that observed and a mean heating of 98 W m^{-2}
14 compared with the observed value of 64 W m^{-2} , due to an overestimation of surface downward
15 shortwave in ERA-I. Differences between TOA and surface fluxes are used to infer atmospheric
16 radiative heating. Effects from TCWV on these are significant, but they are more strongly controlled
17 by AODs, since dust has a much greater effect on surface net radiation than TOA net radiation, while
18 effects of TCWV on TOA and surface heating are more similar.

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

29 Improved modelling of the energy budget of the SHL region is needed in models to improve
30 predictions of the WAM across time scales (e.g. Evan et al., 2015). The results show that it is
31 important that models used for predictions can accurately capture the processes controlling the water
32 vapour distribution over the Sahara, as well as the dust. This capability is currently questionable for
33 both water [Marshall et al. 2013b; Birch et al. 2014; Garcia-Carreras et al. 2013; Roberts et al., 2014],
34 clouds (Roehrig et al. 2013; Stein et al., 2015) and dust [Evan et al., 2014], with many dust errors

1 coming from moist convection [Marsham et al., 2011; Heinold et al 2013]. The results presented
2 here therefore strongly motivate the need to improve the representation of advection of water vapour,
3 clouds and convection in models.

4

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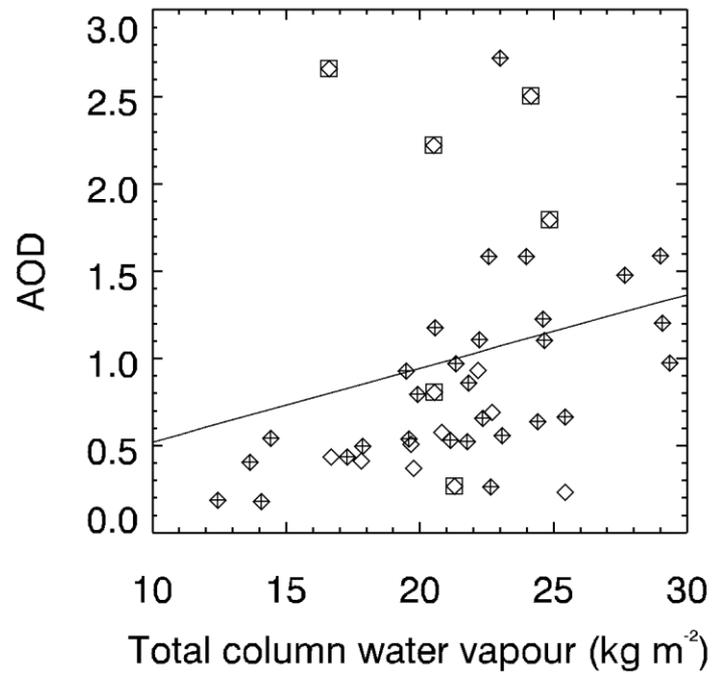
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	Observations		ERA-I	
	Good surface data	All data	Good surface data	All data
TCWV : AOD (kg m ⁻²)	0.04 (0.29)	0.04 (0.30)	0.02 (0.16)	0.02 (0.14)
AOD : TOA Net (W m ⁻²)	5.3 [3.6] (0.26)	3.5 [2.3] (0.17)	1.7 (0.12)	0.33 (0.02)
AOD : TOA Net LW (W m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

1 Table 1. Gradients of best-fit straight lines (i.e. regression coefficients) for listed relationships,
2 values in [] are normalised by standard deviation of TCWV or AOD. Values in () are correlation
3 coefficients (bold values are significant at 90% level). For ERA-I observed AODs are used.
4 Standard deviation in TCWV in ERA-I = 4.7 kg m⁻² (4.5 for “All data”). For observations 4.7 kg m⁻²
5 (4.4 for “All data”). Standard deviation in AOD for observations is 0.68 (0.65 in “All data”).



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2 Figure 1. Daily means TCWV and AOD, pluses show days with complete surface data, squares
3 days with some interpolation (see Section 2). Diamonds show all data points (including days with
4 no surface-flux data).

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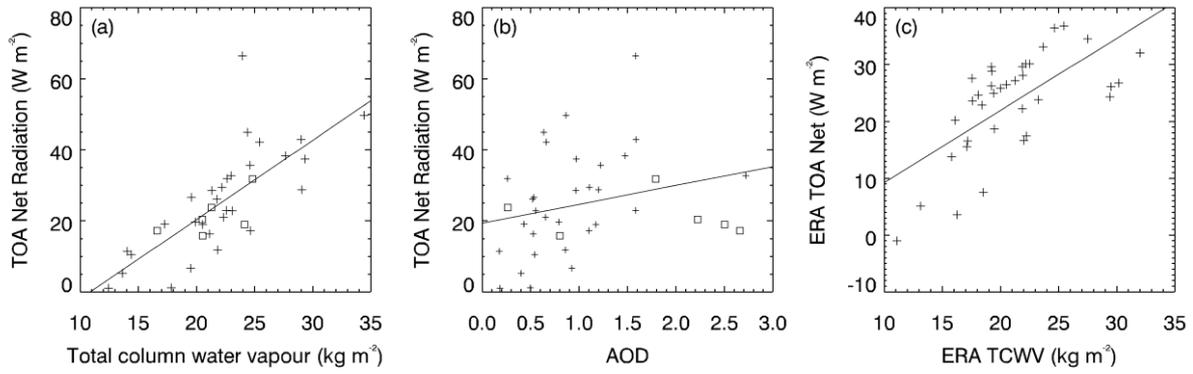
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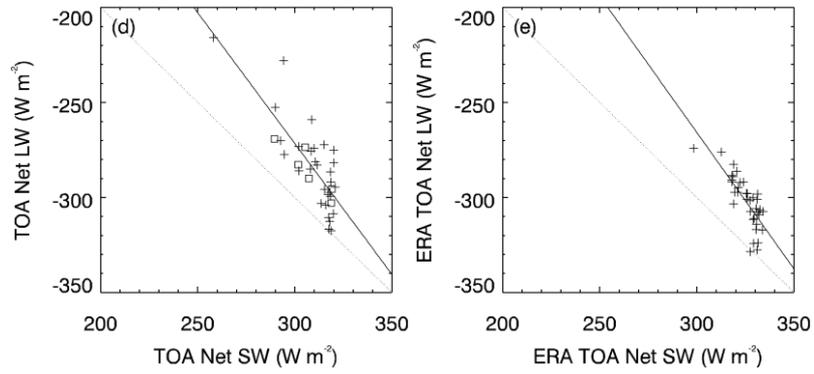
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4 Figure 2. TOA fluxes, with symbols as in Figure 1, showing daily-means for days with surface data.

5 Dotted lines in (d) and (e) have a gradient of -1.

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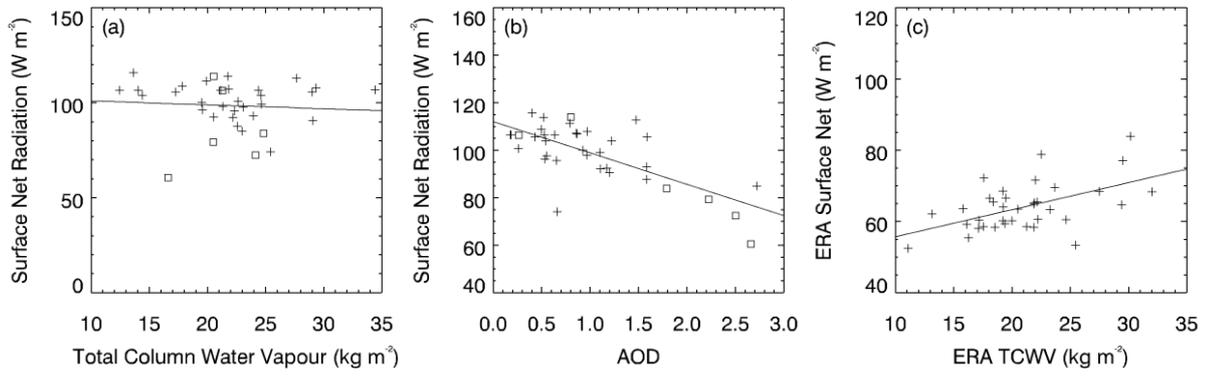
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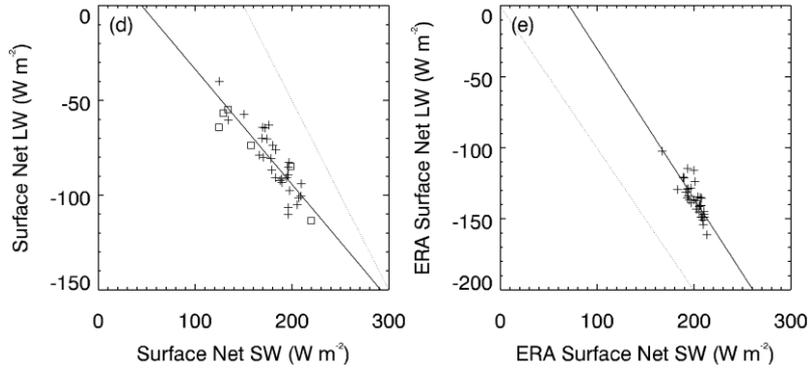
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4 Figure 3. As Figure 2, but for surface fluxes.

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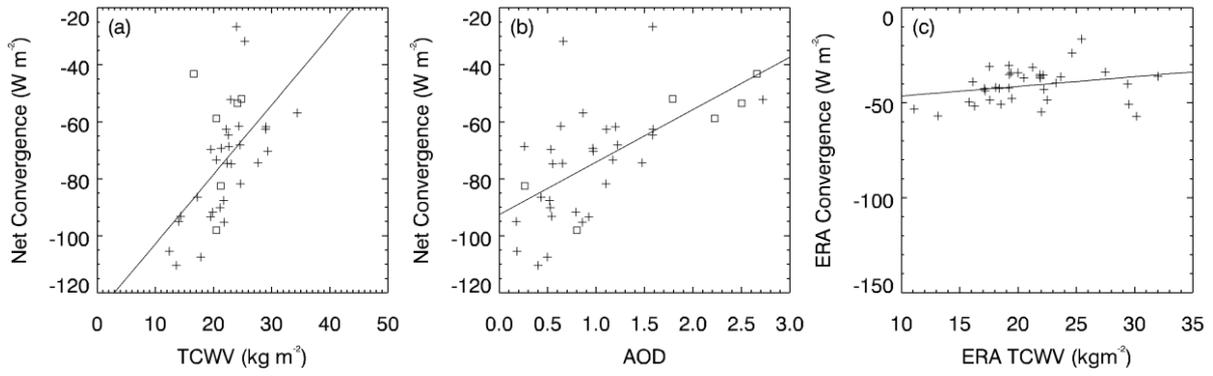
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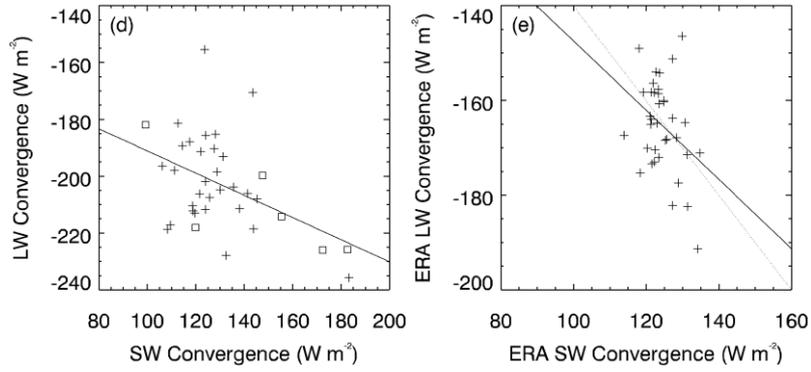
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3 Figure 4. As Figure 2, but for inferred atmospheric heating (TOA flux minus surface flux).

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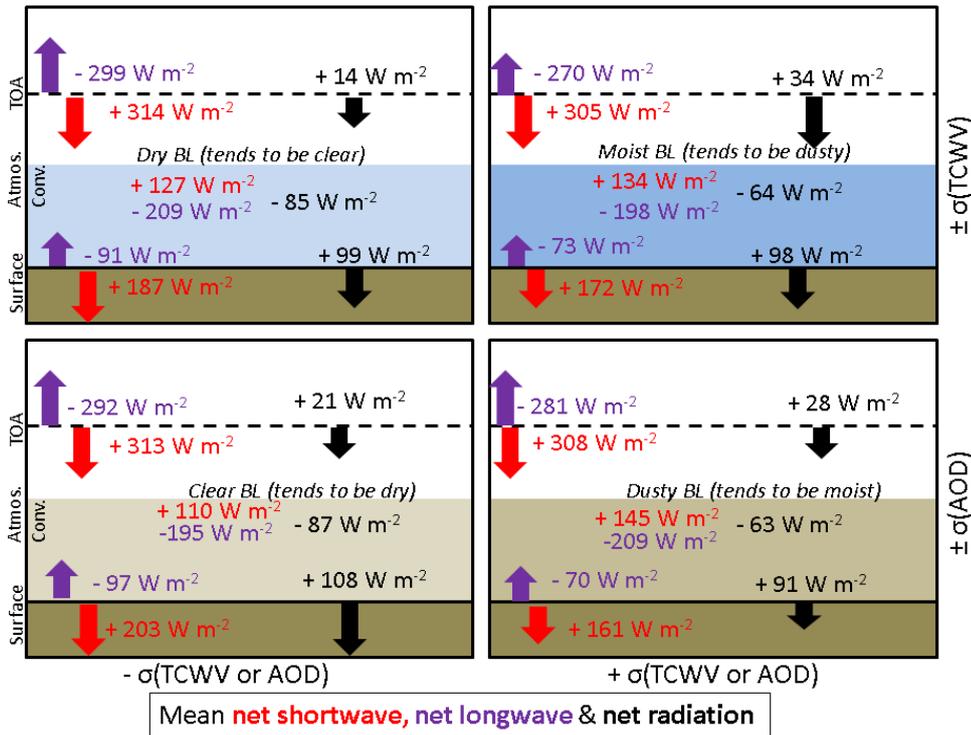
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2 Figure 5. Schematic showing net radiation and implied tropospheric radiative heating, in situations
 3 where either TCWV (top row) or AOD (bottom row) is perturbed by plus or minus one standard
 4 deviation away from their mean state (right and left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

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