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A Study of the Relations between Soil Moisture, Soil Temperatures and Surface Temperatures Using ARM Observations and Offline CLM4 Simulations

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External Editor: Nicole Mölders

Received: 10 January 2014; in revised form: 3 September 2014 / Accepted: 5 September 2014 / Published: 29 September 2014

Abstract: Soil temperature, soil moisture, skin temperature and 2-m air temperature are examined from both ground observations and the offline community land model (CLM4). Two-layer soil moisture and three-layer soil temperature observations from six-year (2003–2008) ground measurements at the Lamont, Oklahoma site supported by the Atmospheric Radiation Measurement (ARM) Program of the Department of Energy (DOE) show clear vertical and temporal relations between soil temperature and soil moisture with surface skin temperature and 2-m air temperature. First, daily means reveal that all of these variables have clear seasonal variations, with temperatures peaking in summer and minimizing in winter as a result of surface insolation. Nevertheless, the 2-m air temperature and upper soil temperature (−0.05 m) peak at 2 h after that of surface skin temperature because of the lag of transport of heat from the skin level to the 2-m air and to underground respectively. As a result of such lag, at the monthly annual cycle scale, 2-m air temperature has higher correlation with upper soil temperature than skin temperature does. Second, there are little diurnal and annual variations at the lowest soil layer (−0.25 m). Third, a negative correlation (−~0.40) between skin temperature and soil moisture is observed, consistent with the expectation that heat flux and evaporation are competing physical processes for redistributing surface net radiation. Soil moisture, however, minimizes in March and maximizes in winter due to the local rainfall cycle. All of these key observed relations are qualitatively reproduced in the offline CLM4 using the

atmosphere forcing derived from ARM observations. Nevertheless, CLM4 is too dry at the upper layer and has less variation at the lower layer than observed. In addition, CLM4 shows stronger correlation between T_{soil} and T_{skin} ($r = 0.96$) than the observations ($r = 0.64$), while the predicted nighttime T_{skin} is 0.5–2 °C higher than the observations.

Keywords: land surface climatology; climate modeling; soil moisture; surface temperature; land surface model; ground observations

1. Introduction

Soil moisture is the quantitative water content in terms of volume (volumetric) or mass (gravimetric) among soil probes. This variable determines various key soil thermodynamic properties, such as heat conductivity, hydroconductivity, and the Bowen ratio. Therefore, soil moisture interacts with soil temperature, skin temperature, and air temperature in controlling the exchange of water and heat energy between the land surface and the atmosphere through evaporation and plant transpiration. Because the land model has a relatively long memory on soil moisture [1], reasonably simulating this variable is essential [2]. However, previous studies have shown that soil moisture has a positive feedback on precipitation patterns [3,4] and various other physical processes such as surface humidity, surface insolation, snowmelt, and runoff [5–8]. In order to improve climate models, the relations between soil moisture and soil temperature need be further studied.

Soil moisture measurement is a challenge, in particular, from satellite measurements and at the global scale. Although remote sensing now can measure many land surface variables, such as vegetation and snow coverage [9–11], soil moisture information is still difficult to obtain from space due to the lack of a convincing remote sensing approach. Consequently, the relations among soil moisture, soil temperature, skin temperature, and air temperature are not well understood [4,12]. Probably at this stage, the most practical approach is to analyze the temporal variations of soil moisture and soil temperature from ground sites, study the vertical distributions of these variables, and reveal their relations with atmosphere and land surface processes.

Six-year instantaneous ground observations of land surface skin temperature (T_{skin}), which is the radiometric temperature derived from surface upward longwave emission, is closely related to land surface radiative properties [13,14]. T_{skin} and 2-m surface air temperature (T_{air}) are both important variables at land surface with different physical meaning and magnitude [15]. T_{skin} , T_{air} , upper soil temperature (T_{soil}) at 0.05 m below ground, middle T_{soil} at 0.15 m below ground, lower T_{soil} at 0.25 m below ground, upper soil moisture and lower soil moisture monitored at the Lamont, Oklahoma Atmospheric Radiation Measurement (ARM) Program of Department of Energy (DOE) site are used to examine the diurnal, seasonal, and vertical relations among these variables. Specifically, the goals of this work are to:

- (a) understand the vertical and temporal variations of volumetric water content;
- (b) understand the relations among soil moisture, soil temperature and surface temperature based on long-duration ground observations;

(c) evaluate the CLM4 simulations with respect to the relations identified in (b).

Section 2 provides data and modeling information. Results are presented in Section 3 in two parts; part 1 is ARM observations and part 2 is offline CLM4 simulations. Final discussions and remarks are given in Section 4.

2. ARM Data and Offline Community Land Model Simulations

2.1. ARM Data

Since 1992, the U.S. Department of Energy (DOE) Atmospheric Radiation Measurement Program (ARM) has built six ARM Climate Research Facilities (ACRF) to continuously measure atmospheric, land, and soil variables. The Lamont, Oklahoma site (36°36'18.0"N, 97°29'6.0"W, altitude 320 m) is one of the ACRF sites analyzed in this paper. Land cover of this site is winter [16] or cropland according to MODIS land cover observation.

Soil moisture and temperature are monitored by the Carbon Dioxide Flux Measurement Systems (CO2FLX, [16]). Volumetric soil moisture content is continuously measured using four replicate soil impedance probes (Delta-T, ML2x) at two depths (−0.05 and −0.25 m, within an effective sampling interval of ~5 cm) with an accuracy of $0.05 \text{ cm}^3 \cdot \text{cm}^{-3}$ estimated by the manufacturer for the case of un-calibrated soils. Soil temperature is measured with four replicate thermocouple probes at three depths (−0.05, −0.15, and −0.25 m) with an accuracy of approximately 0.5 K. It should be noted that the agricultural soil is periodically disturbed by field management, leaving an uneven surface layer that introduces heterogeneity into the placement of the shallow probes. All soil data are averaged to half hour intervals for each depth and sensor type for later analysis. Data was also averaged to daily and monthly in this analysis. 2003–2008 data are used to represent the climatological sense of soil moisture and temperature for that region.

T_{skin} measurement is calculated from surface upwelling (10 m) longwave hemispheric irradiance. Specifically, this data is from the Quality Control of Radiation (QCRAD) value-added product. It has an original temperature resolution of 1 min, and an averaging interval of 1 h.

There were some missing days of data. For example, in 2004, soil temperature was available for 251 of the 366 days. As a result, soil moisture was only compared to T_{skin} on days when both were available (270 for lower, 297 for upper).

2.2. Offline CLM4 Simulations

The Offline Community Land Model (CLM4) [17], which is the land scheme coupled into the Community Earth System Model (CESM) developed at the National Center for Atmospheric Research (NCAR), is examined using ARM-Lamont atmospheric forcing and evaluation data set for year 2004. This paper focuses on evaluating skin temperature, soil temperature and soil moisture analyses because little research has focused on temperature and moisture relations in CLM4. We find that CLM4 simulates skin temperature well in Spring and Fall, but overestimates it in Summer and Winter, by up to 2.8 °C at daily average. In addition, CLM4 is too dry at the upper layer (−0.05 m) and lower layer (−0.25 m). Nevertheless, a large deficiency occurs for snow coverage and consequently albedo simulation in winter.

The CLM4 control simulation is driven by the default offline forcing, which is extracted from NCAR/NCEP reanalysis and provided in CESM model release [18,19], for the one-point (Lamont OK). The sensitivity run (case CLM4-arm-forcing) is forced by the ARM observations taken at the same location in order to reduce the model uncertainties induced by the atmosphere forcing. Specifically, the offline CLM4 is forced using hourly ARM observed direct and diffuse surface solar radiation, wind, precipitation, downward longwave radiation, humidity and 2-m T_{air} . Hourly observation data set (ARM-OK) for land model forcing as well as evaluation is developed from the Oklahoma Lamont site (e.g., CO2flx site) from ARM program. This ARM-OK data set has two parts: forcing data (e.g., ARM-OK-forcing) and evaluation data (e.g., ARM-OK-evaluation). ARM-OK-forcing includes direct and diffusive solar radiation, downward longwave radiation, air temperature, precipitation, wind speed and direction which are needed to force offline land surface model. ARM-OK-evaluation, on the other hand, includes 2-m temperature, T_{skin} , upward longwave radiation, ground flux, two layers of soil temperature, three layers of soil temperature, and CO_2 flux. ARM-OK forcing is used to run offline Community Land). A report on how ARM missing data is treated is currently under review [20]. Furthermore, the offline CLM4 run has a 50 year spin up with soil moisture initialized using ARM observations.

Observations during the year 2004 are randomly selected to force the offline CLM4 simulations and then to evaluate the outputs to demonstrate the performance of CLM4 T_{skin} , soil temperature and soil moisture relations. Other years, such as 2003 and 2005, are also simulated, and the results are consistent with those presented for 2004.

3. Results

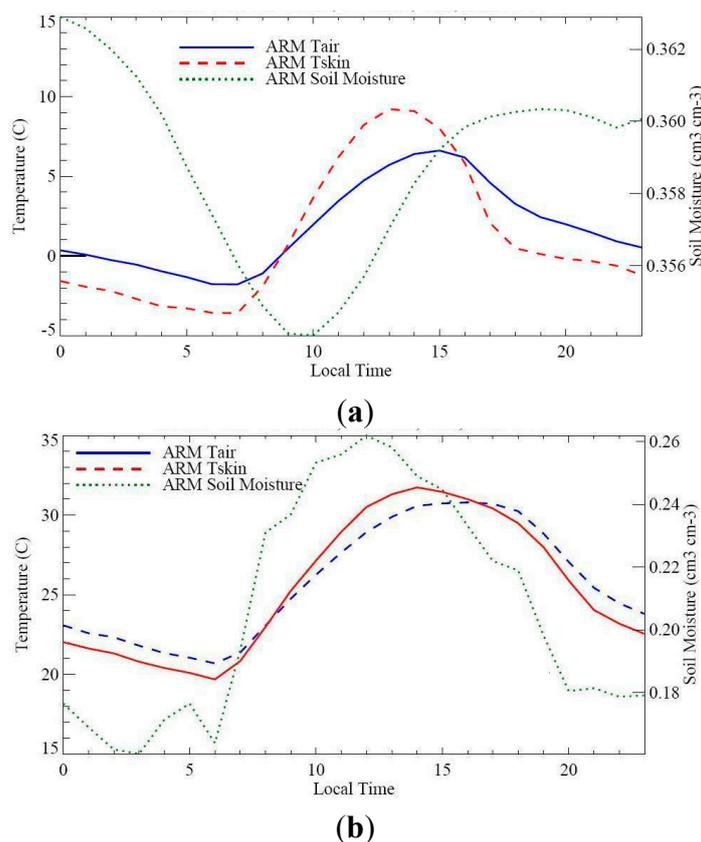
3.1. ARM Observations

The diurnal cycle is evident on upper soil moisture in both January and July 2004 (Figure 1). For July, the soil moisture minimum occurred at 03:00 and maximum is at 12:00, before minimum and maximum temperatures. On the other hand, the January minimum daily soil moisture occurred at 10:00, after the minimum air and skin temperatures. Furthermore, the July soil moisture diurnal cycle is less smooth than January, partly because of some missing days of data for July (28 h of missing data for T_{skin} , 22 h of missing data for T_{air} , 193 h of missing data for the average of upper and lower soil moisture, or about 8 days). Nevertheless, the primary reason for the July cycle being less smooth is precipitation events during this period of time. In addition, the soil moisture is higher in January, with less monthly variation in January, than in July.

Surface T_{skin} and 2-m T_{air} have clear diurnal cycles that follow surface insolation; namely, lowest in the early morning just before sunrise and highest at local noon. Also, due to radiative cooling at night, T_{skin} is lower than T_{air} during the night and then increases due to absorption of ground energy during the day. In addition, T_{skin} is higher than T_{air} during most of the day. A similar pattern also occurs in July with the largest difference between T_{skin} and T_{air} at noon of about 3.0 °C. Such a pattern is typical and is consistent with the earlier observations in other field experiments such as FIFE [13]; namely, T_{skin} is higher than T_{air} during daytime (09:00–16:00) but lower than T_{air} at night and early morning, before the surface heats up by absorbing surface insolation. Other years of ARM observations during

2000–2008 are also examined (not shown) and the temperature patterns are very similar, except that July soil moisture is largely dependent on local time and rainfall occurrence. Jin and Dickinson have proved that such a typical pattern is a function of land cover, latitude, and cloud conditions [14].

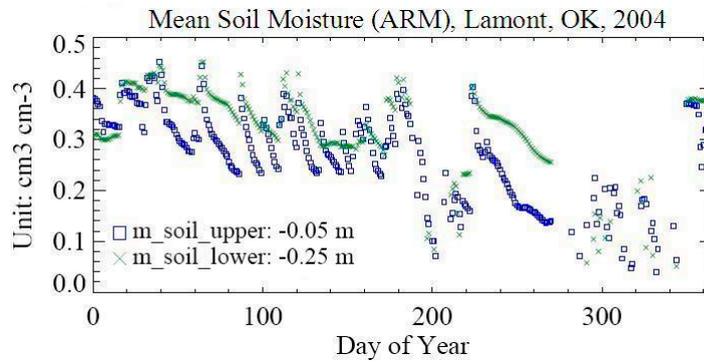
Figure 1. Monthly mean diurnal cycle of skin temperature (T_{skin}), upper-layer soil moisture, and 2-m air temperature (T_{air}) for (a) January 2004; (b) July 2004. Data is from Atmospheric Radiation Measurement (ARM) Lamont, OK site.



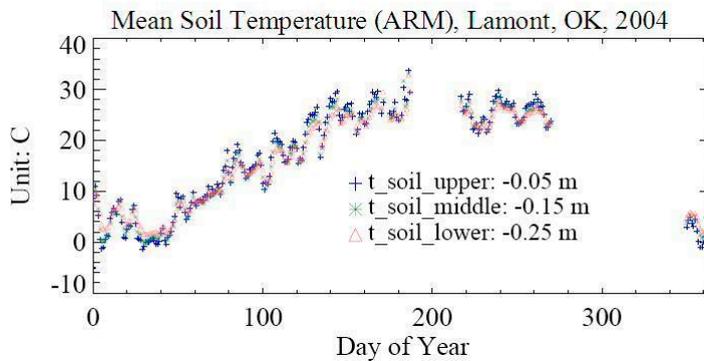
Due to evaporation at the upper soil level (-0.05 m), soil moisture is generally about $0.1\text{--}0.2\text{ cm}^3\cdot\text{cm}^{-3}$ less than that at lower level (-0.25 m, Figure 2a). Nevertheless, soil moisture at the two layers follow each other closely throughout the year with generally high values in winter and spring and low values from July (from Day 190) until the end of fall (Day 300). Nevertheless, the general seasonal trend is frequently interrupted by precipitation events. Winter snowfall and spring snowmelt increases soil moisture, while dry summer conditions lead to greater evaporation and thus reduce soil moisture. Furthermore, in the dry season, both levels have similar soil moisture which is less than $0.25\text{ cm}^3\cdot\text{cm}^{-3}$.

Similar to T_{skin} (shown in Figure 1), T_{soil} at the upper level (-0.05 m), middle level (-0.15 m) and lower level (-0.25 m) have a similar annual cycle, with the peak in summer and minimum in winter following the seasonal cycle of solar radiation (Figure 2b). Over the Lamont site, the minimum soil temperature is about -3.0 °C in December and January, and then almost linearly increases to a July maximum of 32.3 °C. In addition, in summer the upper level soil temperature is higher than that of the lower level due to heat transport from the surface, but an opposite pattern is found in winter due to surface radiative cooling.

Figure 2. (a) Annual variations of daily averaged soil moisture at upper layer (−0.05 m) and lower layer (−0.25 m). (b) Annual variation of daily averaged soil temperature for upper soil layer (−0.05 m), middle soil layer (−0.15 m) and lower soil layer (−0.25 m).

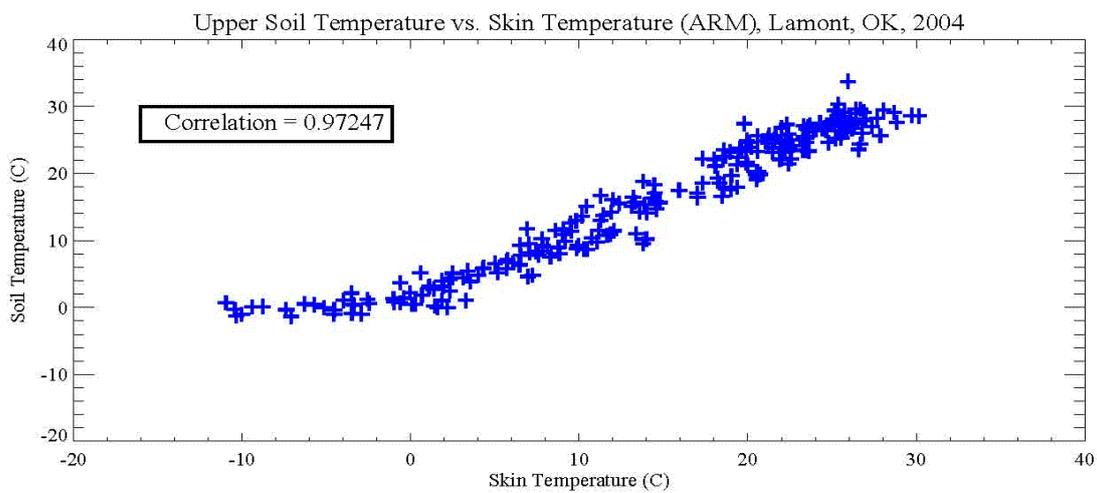


(a)



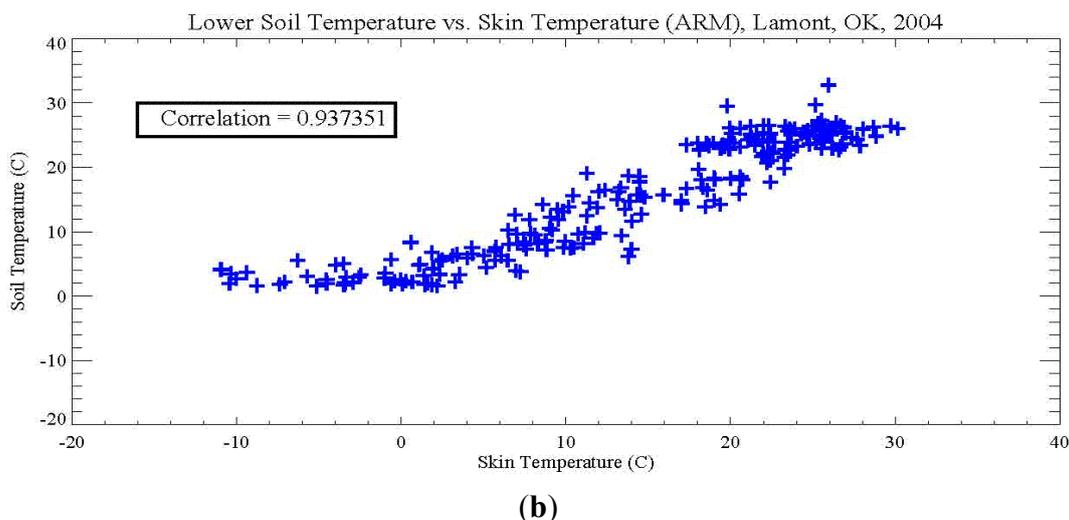
(b)

Figure 3. Soil temperature and skin temperature relations based on annual daily averaged ARM observation at Lamont OK site. (a) upper soil temperature vs. skin temperature, and (b) lower soil temperature vs. skin temperature. Data is daily mean during 2004.



(a)

Figure 3. Cont.



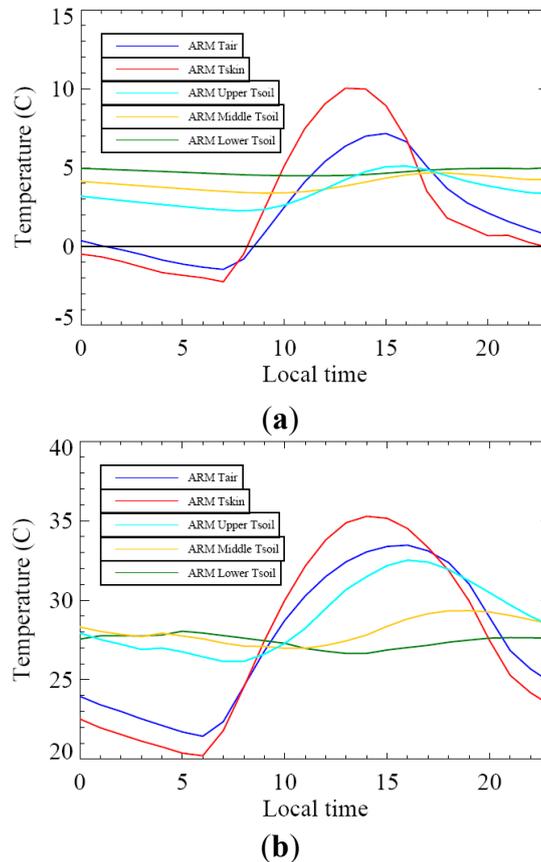
T_{soil} correlates highly with T_{skin} (Figure 3) because of the heat propagation from ground to soil. Sampled for the year 2004, the daily averaged upper T_{soil} and T_{skin} has a high correlation coefficient up to 0.97 (Figure 3a), with a value of 0.94 (Figure 3b) for lower T_{soil} and T_{skin} . This suggests that the upper level T_{soil} is more closely related to T_{skin} because the absorbed surface insolation is partly used to warm the ground and partly transported in terms of ground heat flux to warm soil. More importantly, the close-to-linear relationship between T_{skin} and T_{soil} as well as for T_{air} and T_{soil} may be useful for predicting these two temperatures. Correlation coefficient of 0.97 for daily mean T_{skin} vs. T_{soil} only means that these two parameters are highly related, but it is not linear. Furthermore, the almost flat T_{soil} when T_{skin} is below 0, which means that when the surface is frozen, the soil temperature at particular layer varies little.

In contrast, soil moisture does not correlate well with T_{skin} (not shown). Specifically, a negative correlation coefficient (-0.42 , p -value = 0.05) is detected between the upper level soil moisture and T_{skin} . This low negative correlation coefficient is because SH and LE are competing physical processes to re-distribute the surface absorbed radiative energy. Nevertheless, the low coefficients reveal the loose relationship between these two parameters.

The phase lag in the diurnal variations of temperature (Figure 4) at different layers—2-m T_{air} , T_{skin} , and T_{soil} at upper, middle, and lower levels, respectively, show evident energy propagation among air-land-soil. In January (Figure 4a), T_{skin} has the largest diurnal variation because the skin level (e.g., ground at this location) absorbs surface insolation most rapidly. T_{air} has a smaller diurnal range than T_{skin} but a larger diurnal range than the upper, middle, and lower layers T_{soil} . The upper layer T_{soil} has larger variations than the middle and lower layers T_{soil} . In fact, not much diurnal variation occurs at the low soil layer (-0.25 m), implying that heat can be barely transported to this depth. In addition, the phase lag of diurnal variation is evident: T_{skin} reaches its maximum (10.03 °C) at 13:00, then T_{air} reaches its maximum (7.17 °C) 2 h later (15:00) after the land surface heats up, while soil layers reach maximum at 16:00 for the upper layer (5.10 °C), 17:00 for the middle layer (4.66 °C), and 23:00 for the upper layer (4.99 °C). Furthermore, the minima of these temperatures are -1.46 °C (07:00) for T_{air} , -2.24 °C (07:00) for T_{skin} , 2.25 °C (08:00) for the upper layer soil, 3.63 °C (12:00) for the middle layer soil, and 4.48 °C (11:00) for the lower layer soil. Note that in winter, soil layers are warmer than both

air and ground with deeper soils being warmer than upper soils. Radiative cooling near the ground is very clear at night, leading to lower T_{skin} than T_{air} and T_{soil} .

Figure 4. Multi-year (2003–2008) averaged diurnal variations of 2-m air temperature (T_{air}), skin temperature (T_{skin}), and soil temperature at upper level (−0.05 m from ground), Middle level (−0.15 m) and Lower level (−0.25 m) for (a) January and (b) July.



In July (Figure 4b), the diurnal maxima are 33.46 °C (16:00) for T_{air} , 35.28 °C (14:00) for T_{skin} , and 32.52 °C (16:00), 29.34 °C (19:99), and 28.05 °C (05:00) for upper, middle, and lower T_{soil} , respectively. The minimum for the upper layer T_{soil} is 26.15 °C (07:00), while it is 26.98 °C (10:00) and 26.65 °C (13:00) for the middle and lower layers, respectively. In addition, T_{skin} has a larger diurnal variation (15.07 °C) than T_{air} (12.02 °C). More importantly, the upper T_{soil} has much more significant diurnal range (6.37 °C) than in January, and other soil layers show little variation (2.36 °C, 1.0 °C). In fact, the lower T_{soil} shows an inverse pattern, namely, a maximum in the morning and minimum in the afternoon. Furthermore, T_{skin} and T_{air} both reach minimum at the same time (6 AM), with values of 20.21 °C for T_{skin} and 21.44 °C for T_{air} . In July (Figure 4b), T_{soil} at the lower level (−0.25 cm) still has diurnal variation due to the arrival of heat propagated from surface. Nevertheless, it takes about 10 h for heat to reach this level, and thus T_{soil} of this layer has a clear opposite phase than T_{skin} diurnal cycle. In January (Figure 4a), however, little heat can be transported into this level due to frozen ground, and thus no clear diurnal cycle at this soil layer.

Based on the 6-year (2003–2008) annual diurnal cycle, T_{air} is more closely related to the upper layer T_{soil} than T_{skin} (Table 1). For example, in January, the correlation coefficient between T_{air} and upper T_{soil} is 0.87 and only 0.64 for T_{skin} vs. upper T_{soil} . Similarly, T_{air} vs. the middle layer T_{soil} has a higher

correlation coefficient (0.41) than that between T_{skin} vs. the middle layer T_{soil} (0.07). On the contrary, the lower layer T_{soil} has a higher but negative correlation with T_{skin} . Specifically, the coefficient between T_{skin} and the lower layer T_{soil} in January is -0.53 but -0.26 between T_{air} and the lower layer T_{soil} for the same month. More importantly, seasonality is very clear: In July, both T_{air} and T_{skin} have a much higher correlation with the lower layer T_{soil} than in January. While in July, T_{skin} vs. lower T_{soil} is -0.93 and for January it is only -0.54 . This is probably because lower T_{soil} has little diurnal variation, but the minimum value is in the afternoon, and maximum in the morning (opposite of normal diurnal cycle, as noticed in Figure 4b).

Table 1. Correlation Coefficients based on 2003–2008 monthly mean observations. The correlation coefficients are based on hourly-averaged temperatures for each day of the month between 2003–2008 ($n = 4644$).

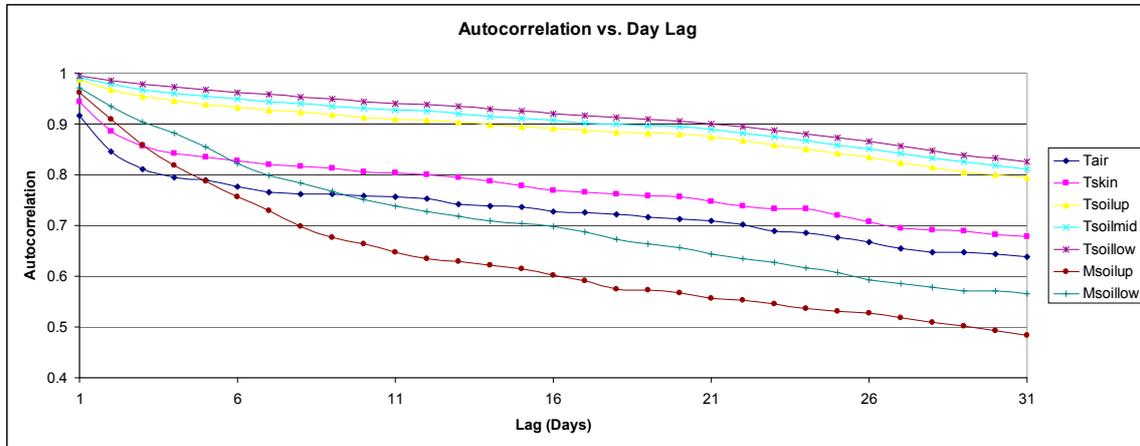
	T_{air} January	T_{skin} January	T_{air} July	T_{skin} July
Upper T_{soil}	0.8707	0.6431	0.8842	0.8023
Middle T_{soil}	0.4165	0.0753	0.2978	0.1231
Lower T_{soil}	-0.2614	-0.5396	-0.8795	-0.9382
T_{skin}	0.9327	-	0.9831	-

Note that T_{skin} and upper T_{soil} have a 0.97 correlation (Figure 3a) while it is only 0.80 for July and 0.63 for January in Table 1. This is probably because Table 1 is based on the multi-year averaged diurnal cycle for 2003–2008, namely, an average over all days to get a diurnal cycle for the month, and then averaged over all years. Furthermore, Table 1 represents the inter-annual variations of the variable.

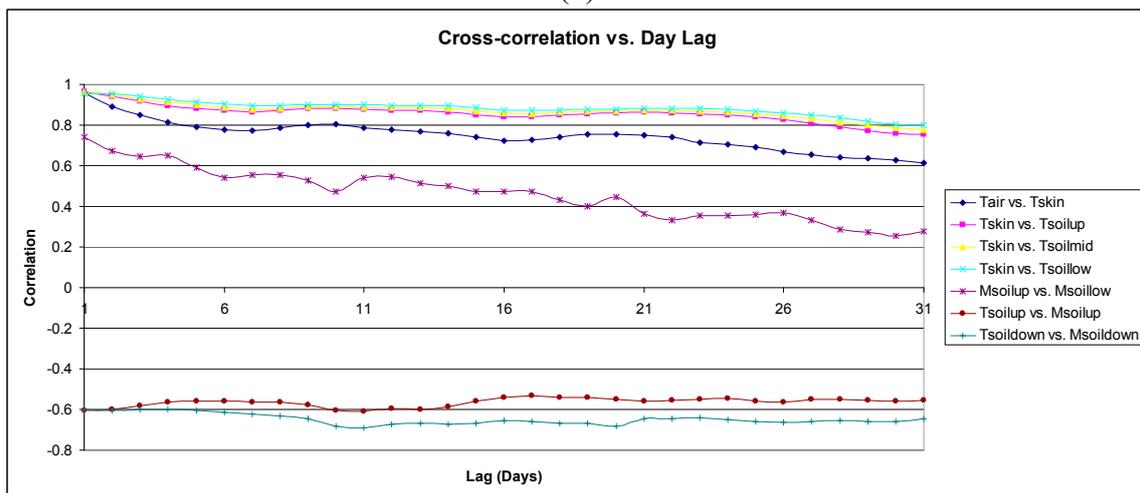
Autocorrelation presents self-persistence of a variable. Soil temperatures have the highest autocorrelations with day lag (Figure 5a). The deeper the soil, the greater the self-persistence is as reflected in the autocorrelation coefficients. In addition, the lag autocorrelation for soil moisture decreases faster than other variables, implying the least self-persistence among these parameters. Furthermore, the upper layer soil moisture has less self consistency than the lower soil moisture after one day.

Cross-correlation with day lag (Figure 5b) shows that T_{skin} and the upper layer T_{soil} have the highest cross-correlation (~ 0.90), and that cross-correlation increases with depth, namely, higher correlation of T_{skin} with the middle and lower layers T_{soil} than T_{skin} with the upper layer T_{soil} . In addition, T_{skin} and T_{air} have relatively low cross-correlations, from 0.96 at a day lag of 1 dropping to 0.8 at a day lag of 5. Furthermore, the upper layer soil moisture and the lower layer soil moisture have the lowest positive cross-correlation, and this value rapidly decreases as day lag increases. Again, T_{soil} and soil moisture have a negative cross-correlation, of around -0.6 at a day of lag 1 and remaining mostly unchanged as day lag increases.

Figure 5. (a) Autocorrelation with day lag using daily averaged data from 2003–2008 data. T_{air} is 2-m air temperature. T_{skin} is skin temperature. T_{soilup} is upper soil temperature. $T_{soilmid}$ is middle layer soil temperature. $T_{soillow}$ is lower layer soil temperature. M_{soilup} and $M_{soillow}$ are upper and low soil moisture, respectively. (b) Cross-correlation with day lag.



(a)



(b)

3.2. CLM4 Simulations

Soil moisture simulations in CLM4 do not get the absolute values correct, but well simulate the peaks, partly because the model responds to rainfall in the atmospheric forcing data (Figure 6). Specifically, in the model layer 3 (comparable to ARM upper layer -0.05 m, Figure 6a), soil moisture values are systematically lower than the observations, revealing that the model is too dry at this layer. Once rainfall events occur, however, the soil moisture in the model is close to observations as a result of the model’s response to rainfall forcing. From then, the model soil moisture drops quickly from rainfall-induced peaks and the modeled values are apart from the observations by about $0.10 \text{ mm}^3 \cdot \text{mm}^{-3}$ (e.g., $0.40 \text{ mm}^3 \cdot \text{mm}^{-3}$ in day 50 but $0.30 \text{ mm}^3 \cdot \text{mm}^{-3}$ in CLM4). This suggests that the model loses soil moisture too quickly at this layer. Further analysis implies that one possible reason for this is the large latent heat flux in the model at the ground which reduces soil moisture too quickly from lower soil layers. The averaged root-mean-square error (rms) is $0.04 \text{ mm}^3 \cdot \text{mm}^{-3}$ and the

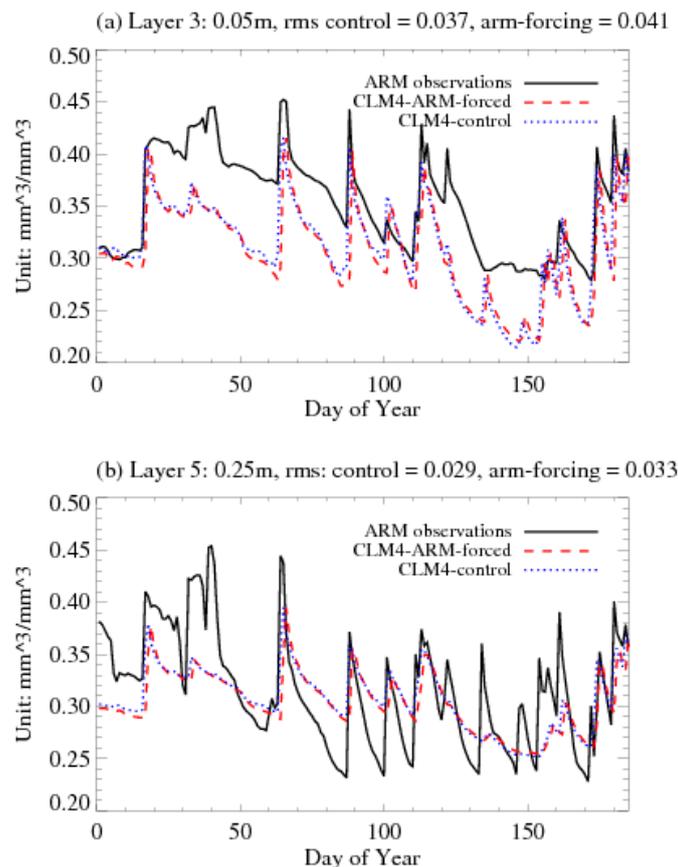
ARM-forced run has relatively larger rms than the control run does, which is unexpected because the ARM-forced run has soil moisture spin up, accurate solar radiation, and rainfall from observations in the forcing. At the lower layer (−0.25 m, comparable to model layer 5, Figure 6b), CLM4 simulated soil moisture is lower than the observations in spring (Julian days 1–120); therefore, the model is still dryer than the observations during this period of time. However, during summer (Julian days 121–180), model simulated soil moisture in general is higher than the observations. Namely, model lower layer is wetter than the observations in summer. In addition, one problem identified is that the model soil moisture is lower than the observation from the very first day (day 0). Since the CLM4 ARM-forced run has soil moisture spin up, the modeled soil moisture should be close to the ARM observation at least at the beginning as day 0–5 in layer 3 (Figure 6a). Further analysis reveals that while CLM4 has default saturation soil moisture and even spin-up soil moisture is close to the observation, the model replaces the spin up soil moisture using the default saturation value. The problem may also explain why, soon after a rainfall event, CLM4 soil moisture increases to close to observation value, but keeps dropping to $0.30 \text{ mm}^3 \cdot \text{mm}^{-3}$. Note that Figure 6 only shows spring and summer because in fall and winter, soil moisture is frozen and causes problems in the ARM observations. Furthermore, the very similar results in the control run and the ARM-forced run suggest that the model has a problem in its soil moisture. Nevertheless, when the forcing has rainfall, it brings the modeled soil moisture value close to the observations. This means that the forcing is important. The better the forcing, the better the model results.

Examining the correlation of soil moisture between the upper layer and lower layers reveals the dependency and vertical structure of soil moisture. In ARM, the correlation coefficient of soil moisture at these two layers is only 0.64 (Figure 7a) while in CLM4, this correlation coefficient is 0.96 (Figure 7b). Therefore, ARM data are very scattered (Figure 7a) but CLM values are very close to each other (Figure 7b). In addition, the range of soil moisture for layer 5 is $0.30\text{--}0.48 \text{ mm}^3 \cdot \text{mm}^{-3}$ in ARM data while it is only $0.20\text{--}0.42 \text{ mm}^3 \cdot \text{mm}^{-3}$ in CLM4. In conclusion, two weaknesses are identified here for CLM4: First, in the model, in terms of soil moisture, model layers are too tightly coupled; and second, at the lower layer, soil moisture is too low and the range is too small ($0.25\text{--}0.40 \text{ mm}^3 \cdot \text{mm}^{-3}$ in CLM4 while ARM data is $0.18 \text{ mm}^3 \cdot \text{mm}^{-3}$).

Similar to Figure 3, CLM4 shows the high correlation between soil temperature and T_{skin} (Figure 8). The correlation coefficients on daily averaged soil temperature and T_{skin} are 0.99 for upper layer and 0.97 for lower layer. Comparing this with the ARM observations where correlation coefficients are 0.97 for the upper layer and 0.94 for the lower layer (Figure 3), CLM4 has slightly higher dependence between soil temperature and T_{skin} than the ARM observations. The overall relation is well simulated by the model.

Similar to the ARM observations (Figure 2b), soil temperatures in CLM4 Layers 3, 4 and 5 are very close to each other (Figure 9). Apparently, all three layers have the same seasonality because the surface insolation is the key factor determining soil temperature. Nevertheless, layer 3 (−0.05 m) has larger temperature variations than the other two underlying soil layers, as expected. For example, around Julian day 40, T_{soil} for layer 3 is 2–3 °C lower than that of layer 4 and 5, but around Julian day 200, it is higher than the other two layers by about 2 °C. Such larger seasonality and daily variation are partly due to stronger ground heat flux transport between surface and the upper soil layer and partly due to the fact that model is too dry at this layer, as revealed in Figure 6a.

Figure 6. Simulated daily averaged volumetric soil moisture from offline community land model (CLM4) in (a) the upper layer (layer 3, -0.05 m) and (b) the lower layer (Layer 5, -0.3 m). The control run (CLM4-control) is forced by the default atmospheric forcing provided by National Center for Atmospheric Research (NCAR) model release package and the ARM-forced run (CLM4-ARM-forced) uses the hourly ARM observed atmospheric forcing data in Year 2004.



The diurnal phase of 2-m T_{air} , T_{skin} , and T_{soil} are well simulated in offline CLM4 (Figure 10), comparing those to ARM observations (Figure 4). First, T_{skin} is lower than T_{air} at night but warmer than T_{air} during daytime after sunrise, a feature consistent with many previously studies [15]. Second, T_{soil} in layer 3 is lower than T_{skin} with 3 h time lag because heat is propagated from the surface skin layer into the ground. Furthermore, the lower layer T_{soil} (-0.25 m) has the opposite phase to T_{skin} and T_{air} in both January and July, namely, high at local night but low at local daytime. Another deficiency in the model is that $T_{\text{soil}3}$ is higher than T_{air} in July 2004 (Figure 10b), which is not found in the ARM data (Figure 4b). Nevertheless, the most evident deficiency between the CLM4 and the ARM observations occur at nighttime on T_{skin} and T_{air} . In the real world (Figure 4), T_{skin} is lower than T_{air} by about 0.5 °C to 2 °C in January and July, respectively. However, in the model, these two temperatures are too close at nighttime because T_{skin} is too high. Since in an offline model run, T_{air} , downward longwave radiation, and wind are all from the observed ARM observations, this nighttime deficiency suggests some model problems which lead to a hot skin surface in CLM4. In addition, the diurnal scale peak of July T_{skin} is higher in CLM4 than in the observations.

Figure 7. Correlation of volumetric soil moisture between the upper and lower layers: (a) is the ARM observations and (b) is the CLM simulations. Data is daily average for year 2004.

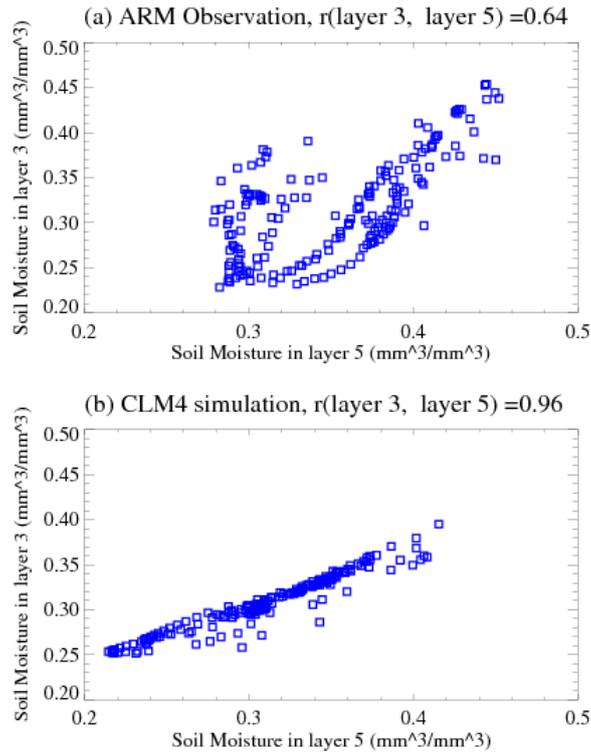


Figure 8. Same as Figure 3, except that skin temperature and soil moisture are from offline CLM4 ARM-forced run for daily averaged value in Year 2004, (a) the upper soil layer (−0.05 m) and (b) the lower soil layer (−0.25 m).

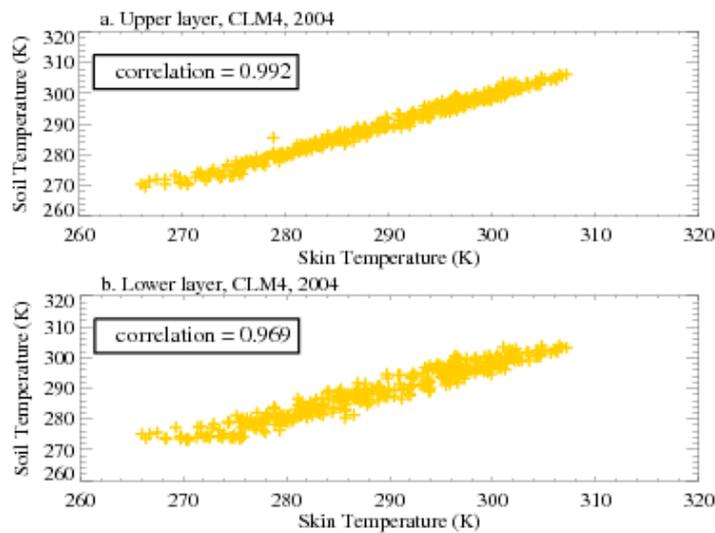


Figure 9. Offline CLM4 simulated daily averaged soil temperature at the upper soil layer (model layer 3, -0.05 m), the middle layer (model layer 4, -0.12 m) and the lower layer (model layer 5, -0.25 m). The model run uses ARM hourly atmospheric forcing for Year 2004.

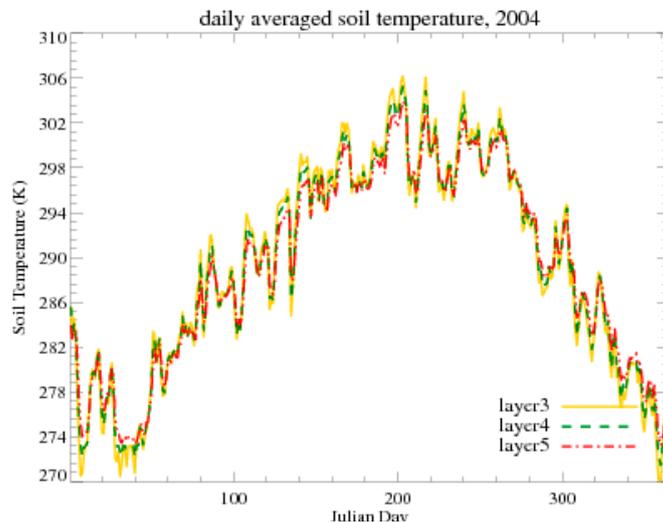
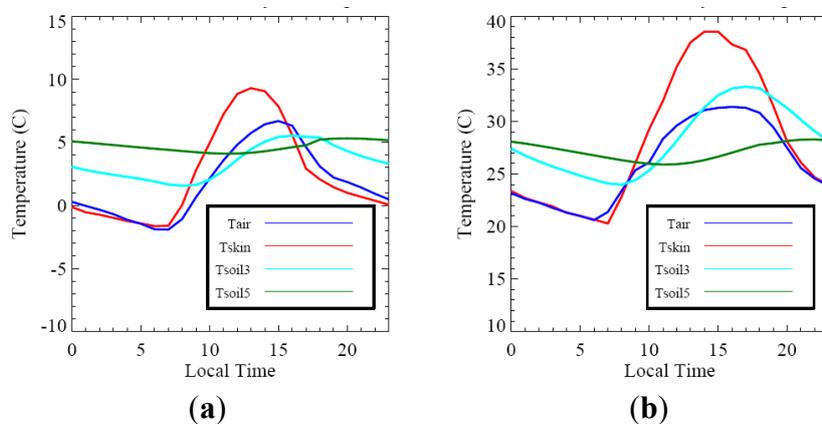


Figure 10. CLM4 simulated monthly mean T_{skin} , T_{air} , T_{soil} at the upper soil layer (T_{soil3}) and T_{soil} at the lower soil layer (T_{soil5}) for (a) January 2004 and (b) July 2004. The model run is forced by ARM hourly observations.



4. Final Remarks

As the key parameters in a land surface model, vertical distribution of soil moisture and soil temperature show evident relations between skin and air temperatures. New findings of this work include:

- At different scales, the signals of these variables are different. For example, at the daily scale, T_{skin} and T_{soil} have high correlation coefficients of up to 0.97. However, on an annual diurnal cycle scale, the correlation between T_{skin} and upper T_{soil} correlation is only 0.64 for January and 0.84 for July. This means that the features of T_{skin} , T_{air} , T_{soil} , and soil moisture need very high temporal monitoring in order to realistically understand the variations and mechanisms of these changes.

- b. The lower layer T_{soil} (-0.25 m) has a negative correlation with T_{skin} and T_{air} at the diurnal cycle scale. Specifically, at -0.25 m of this specific location, T_{soil} does not vary much with the seasonality of surface insolation. CLM4 reproduces these correctly.
- c. T_{air} has a higher correlation with the upper soil layer T_{soil} because both T_{air} and upper T_{soil} need time to be heated up after T_{skin} increases. Therefore, the thermal inertial at the air layer and upper soil layer needs to be considered in the model to reproduce the time lag in heating up. CLM4 simulates this feature well.
- d. T_{soil} has higher self-consistency than T_{skin} and T_{air} . T_{skin} has a slightly higher self-consistency than T_{air} . This may also be explained by the different heat capacity of soil, surface, and air.
- e. Soil moisture is closely controlled by precipitation events. Soil moisture and T_{skin} have a negative correlation since more soil moisture leads to more absorbed radiative energy redistributed into latent heat flux instead of sensible heat flux.
- f. Although CLM4 is capable of simulating the relations of T_{skin} , T_{soil} , and moisture reasonably well, a few critical deficiencies are identified: First, CLM4 is too dry at both upper and lower soil layers. Second, CLM4 shows too much dependency between moisture on different layers. Third, nighttime T_{skin} is 0.5 – 2.0 °C higher than the observations. This nighttime deficiency suggests some model problems which lead to a hot skin surface in CLM4. However, to be more explicit: further research is needed in order to determine whether the error is in the soil transfer or upward longwave or emissivity.

One important finding from this work is that high quality, high temporal resolution observations are needed to understand the highly heterogeneous parameters of the land surface soil moisture. In addition, good atmospheric forcing is critical in order to evaluate the land model in offline mode.

Acknowledgements

This work is supported by NSF larger-scale dynamics program. We thank Marc Fischer for discussion on his CO2flx data and editing the manuscript. The data used in this study were collected by the Atmospheric Radiation Measurement Program Southern Great Plains facility, operated by the Office of Biological and Environment research within the U.S. Department of Energy.

Author Contributions

The authors contributed equally to this work.

Conflicts of Interest

The authors declare no conflict of interest.

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