Dear Anonymous Reviewer 1:

We greatly appreciate for your insightful comments and suggestions. We have revised our paper thoroughly in light of your comments. The changes have been highlighted in color in the manuscript, and also they are summarized below:

1. We have added the explanation of instantaneous observations at Paragraph 2 Section 2.5;

2. We have added the comments about significance level of aerosol, cloud, water vapor and related radiative fluxes variations in related explanation of figures;

3. The details of products used in this study have been added as the Table 1. Moreover, the analysis of uncertainty about the products have added at the introduction of every products in Section 2;

4. We defined the study domain within the South Asian subcontinent to restrict the effect of the Tibet Plateau and sea pixels on the results of this study (Figure 1);

5. We have added the reason about focusing on the monsoon season at Paragraph 2 of Section 2.5. Moreover, we have compared the results between monsoon and non-monsoon seasons, and added the related descriptions in Section 3.1.

Please refer to the following point-to-point responses for more details.

Thank you for your time.

Sincerely,

Dr. Zengxin PAN
General Comments:

The authors examine changes in surface solar radiation (SSR) in the region of India during the monsoon period (June to September) from 2006 to 2015. They relate the SSR changes to changes in aerosol optical depth (AOD), in clouds, and in water vapor. The study relies mostly on satellite products, notably CloudSat, CALIPSO, CERES, and MODIS. The authors find SSR to increase with time, which they ascribe to a decrease in clouds that overcompensate the effect of increasing AOD.

The overall topic of the study - changing SSR in India and its causes - is of interest and suitable for ACP. However, in its present form the study suffers from various shortcomings and a general lack of precision, as detailed below under major (and minor) comments. In particular, the data presented looks, at least in parts, as if it were not a June-September average but rather representative for a specific time of the day. This casts doubt on the entire analysis. Therefore, I recommend rejection of the manuscript in its present form.

Major Comments:

1. SSR magnitude / averaging period. The study deals (see its title) with recent SSR brightening in India. Looking at figure 1b, which shows corresponding key quantities (short-wave radiation at TOA and surface), two thoughts / questions come to mind. First, the radiative fluxes shown are very high, e.g. "TOA Incoming Solar" is around 1230 W m-2 whereas one would expect for a June mean for India something more on the order of 400 to 500 W m-2. Why? See also the next two major comments. Second, the SSR trend shown in the figure has a p-value of 0.42, thus the SSR brightening the authors aim to explain is not statistically significant.

RE: Thank you for your suggestion.

(1) Why the radiative fluxes shown are very high?

In this study, all data indicate observed instantaneous information. We use the data of CloudSat, CALIPSO, MODIS and CERES in daytime. This satellites/sensors cross around equator in daytime at 13:30 local time, at which the incoming solar radiation is almost maximum within the day. Therefore, the instantaneous observation of radiative fluxes provided by CERES are much more than the daily average.

We have added the corresponding explanation at Paragraph 2 Section 2.5 as
the following:

Additionally, the information about aerosol, cloud, and water vapor is calculated on the basis of instantaneous observations, which we then use to ensure consistent conditions for the observations of multiple satellites. The satellites or sensors of the A-Train cross the equator in the daytime at 13:30 local time, at which point the incoming solar radiation is maximum within the day (Liou, 2002). Therefore, the instantaneous observations of radiative fluxes used in this study are greater than the daily average. The results are more representative of the atmospheric condition variations at nominally 13:30 local time than the daily average.

(2) About the statistically significance.

In this study, the statistically significance ($p$-value) of radiative fluxes can be contributed by many factors: (1) the time series of data in this study is just ten years. The confidence of result in showing the temporal variation is highly and positively correlated with the data span; (2) the observations of polar orbit satellite are significantly affected by the widespread and various cloud in monsoon season.

Actually, we can get the aerosol, cloud and water vapor variations with high significance. The aerosol loading quickly increased in recent decades (Suresh et al., 2013; Srivastava, 2016). Moreover, the recent weakened rainfall and dried monsoon during summer monsoon season have been verified by many previous studies (Bollasina et al., 2011; Turner and Annamalai, 2012; Annamalai et al., 2013). The decreases in water vapor amount and clouds significantly contributed to the brightening, further affecting the surface warming in the South Asia. However, the resultant radiative impact of aerosol, cloud and water vapor variations are complicated and fluctuated during the monsoon season based on the observation of CERES. This phenomenon just confirm the fluctuated and unstable changes in the transition period, which is consistent with the study of Soni et al. (2016).

Furthermore, because the sun is the only significant energy source for the global ecosystem, any change in this precious energy source affects our habitats profoundly (Wild, 2012). The findings of this study is considered suggestive to focus on the recent change of solar radiation and its cause. In addition, this study is a new research perspective for understanding the
brightening during the South Asian monsoon season, and more data will be 
acquainted and used to enhance the statistically significance in our future 
study.

2. A-train data. A-train has a small footprint, what then is the overpass statistics over 
domain of interest? What is the statistical uncertainty of an individual June to 
September data point, e.g. CloudSat vertical cloud frequency? Is overpass time 
(A-train equator crossing around 1:30 am and 1:30 pm local time) an issue for the 
presented analysis? The "TOA Incoming Solar" in Figure 1b may suggest so, as 
its value is around 1230 W m-2, whereas a June mean for India is more on the 
order of 400 to 500 W m-2. If the data is instantaneous, how representative is it 
for the entire monsoon season?

RE: Thank you for your suggestion. We have added the explanation about how to 
calculate the spatial and temporal average parameter of aerosol, cloud and water 
vapor and their radiative effect at Paragraph 2 of Section 2.5 as the following:

For the method of spatial and temporal averages, we firstly obtain the spatial 
distribution of temporal average within a 1° × 1° grid during the monsoon season. 
We then determine the spatial average on the basis of the above temporal average 
in the entire South Asian subcontinent. For the spatial patterns of temporal 
changes, we identify the results of the linear temporal changes in atmospheric 
conditions during the monsoon season from 2006 to 2015.

We have added the corresponding explanation about the uncertainty of CloudSat 
at Paragraph 2 and 3 in Section 2.1 and 2.5, respectively, as the following:

**Paragraph 2 in Section 2.1:**
Considering the insensitivity of CloudSat to thin cirrus, Austin et al. (2009) 
showed that 2B-CWC-RO produces typical biases in IWC of −40% to +25% 
versus the observationally derived synthetic data. However, L'Ecuyer et al. (2008) 
proposed that the impact of the thin cirrus that is not detected by CloudSat on 
shortwave (SW) radiation at the surface is −1.2 W m⁻² in a globally averaged 
experiment on radiative fluxes; this value is much smaller than the impact value 
of low clouds.

**Paragraph 3 in Section 2.5:**
We define the cloudy-sk condition for all products as the scenes in which the 
cloud mask from the CloudSat footprint exceeds 20, and the others are defined as
clear-sky condition. Under this screened method of cloud, the probability of a false detection to CloudSat is less than 5% in comparison with that by CALIPSO (Mace and Zhang, 2014).

Moreover, A-Train satellites cross around equator in daytime at 13:30 local time, at which the incoming solar radiation is maximum within the day. The results in this study indicate the atmospheric variations in nominally constant solar zenith angle. The FIG 1 of the Response show the spatially average solar zenith angle in the moment of satellite observation during monsoon season in the South Asian subcontinent. The standard deviation of average solar zenith angle is mostly less than 0.4, indicating the interannual robustness of solar zenith angle in the moment of satellite observation (FIG 1b). Therefore, the results of this study were insignificantly affected by the solar zenith angle, and are representative for the atmospheric condition variations at nominally constant solar zenith angle.

![FIG 1 of the Response](image)

The spatial distribution of the average (a) solar zenith angle and its standard deviation at the surface in the moment of satellite observation detected by CERES during monsoon season in the South Asian subcontinent from 2006 to 2015.

We have added the corresponding explanation about the representativeness of the results at Paragraph 3 in Section 4 as the following:

The data in this study were from the instantaneous observations of multiple sensors. Thus, the magnitudes of radiative impact may were overestimated for the water vapor and cloud variations. The results are more representative of the atmospheric condition variations at nominally constant solar zenith angle than the daily average. In sum, the conclusions of this study are beneficial in the
evaluation of the recent variations in aerosol, cloud, and water vapor and their resultant influences on solar radiation.

3. Products used. It would be helpful if the authors stated more clearly, in Sections 2 and 3 as well as in the figure captions, which variable they take from which satellite / model product and what the intrinsic uncertainty is of the product and why. Also, it should be discussed to what degree different products (satellite and model) can be inter-compared / combined.

**RE:** Thank you for your suggestion. We have added the data source of every figure in the figure captions. Also, we have added the corresponding table of products used in this study in Table 1 as the following and highlighted in color:

**Table 1. Observations used in analysis along with their sources and spatial resolutions.**

<table>
<thead>
<tr>
<th>Sensor</th>
<th>Spatial Resolution</th>
<th>Products</th>
<th>Parameter</th>
</tr>
</thead>
<tbody>
<tr>
<td>CloudSat</td>
<td>1.3 × 1.7 km</td>
<td>2B-CLDCLASS</td>
<td>Cloud Fraction; cloud top/base height;</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2B-CWC-RO</td>
<td>Cloud water content, particle number concentration, and particle effective radius</td>
</tr>
<tr>
<td></td>
<td></td>
<td>2B-FLXHR</td>
<td>Vertical heating rate; SW/LW radiative fluxes</td>
</tr>
<tr>
<td>CALIPSO</td>
<td>5 × 5 km</td>
<td>Level 2 Aerosol Layer</td>
<td>Aerosol optical depth; vertical feature mask</td>
</tr>
<tr>
<td>MODIS</td>
<td>1° × 1°</td>
<td>Level 3 MYD08</td>
<td>Aerosol optical depth</td>
</tr>
<tr>
<td>ECMWF</td>
<td>2.5° × 2.5°</td>
<td>ECMWF-AUX</td>
<td>Pressure; temperature; relative humidity</td>
</tr>
<tr>
<td>CERES</td>
<td>20 × 20 km</td>
<td>Level 2 Single Scanner Footprint</td>
<td>SW radiation in TOA and Surface</td>
</tr>
</tbody>
</table>

The analysis of intrinsic uncertainty about the products used in this study have added at the introduction of every products in Section 2. Moreover, the combined method of multiple satellite products have been added at Paragraph 1 Section 2.5 as the following:

CloudSat, CALIPSO, and Aqua in A-Train are maintained in a tight orbital coordination within 176 s (Stephens et al., 2017). This setting allows near-simultaneous observations of a wide variety of atmospheric and surface parameters, facilitating the comparison of satellite between each other, allowing for even more comprehensive studies of climate.

4. Season chosen. The authors focus on the monsoon season, June to September, when AOD tends to be lowest over India and cloud cover tends to be particularly pronounced (e.g. Raja et al., Atmos.Env. 142, (2016), p. 238-250; or Nair et al.,
Clim.Dyn. 49, (2017), p.1411-1428). Why this choice? And how do results for this season compare with, for example, results based on annual means?

**RE:** Thank you for your suggestion. We have read the above two paper, and cite them at Paragraph 2 and 4 of Section 3.1, respectively. We add the reason about focusing on the monsoon season in this study at Paragraph 2 of Section 2.5 as the following:

This study focuses on the effect of aerosol, cloud, and water vapor variations during the monsoon season. Due to the increased convective activity, the changes in aerosol, cloud, and water vapor and their impact on solar radiation possibly occur more obviously during the monsoon season than during non-monsoon seasons (Turner and Annamalai, 2012).

Moreover, we have compared the results between monsoon season and non-monsoon season, and added the description about these results in Section 3.1 as the following:

Figure 2 shows the recent changes in aerosol loading during different seasons, namely, pre-monsoon, monsoon, and dry seasons, respectively. These three seasons occur in the periods of March–May, June–September, and October–February, respectively. The aerosol loading increased, and the change in the average AOD detected by CALIPSO (approximately 0.13) was more intense than that detected by MODIS (approximately 0.05) during the monsoon season in the South Asian subcontinent from 2006 to 2015 (Figure 2a). These phenomena are mainly due to the effect of aerosols under clouds, which can be detected by CALIPSO but missed by MODIS because of MODIS only inferring the aerosol loading under clear-sky condition (Winker et al., 2010). Moreover, CALIPSO can identify aerosol and cloud layer with ~90% accuracy. The sum of dust, smoke, and polluted dust aerosol loadings is close to the total aerosol loading (orange line in Figure 2a). This phenomenon indicates that aerosols mostly consist of dust, smoke, and polluted dust during the monsoon season in the South Asian subcontinent, which is consistent with the previous study (Das et al., 2015). These aerosols all present distinct absorbing effects on solar radiation, thereby attenuating the incoming solar radiation more intensively than other types of aerosols (Logan et al., 2013).

Figure 2b shows the consistent increase in AOD during the pre-monsoon, monsoon, and dry seasons, respectively. The increment of AOD during the
pre-monsoon and dry seasons (by 0.08 and 0.11, respectively) were greater than those during the monsoon season according to the MODIS observations from 2006 to 2015. The significant level of AOD variation detected by MODIS was higher during non-monsoon season ($p<0.01$) than during the monsoon season, which mainly due to the disturbance of widespread clouds during the monsoon season (Remer et al., 2005). The change in the average AOD detected by CALIPSO showed a higher significance level ($p=0.07$) than that detected by MODIS due to the minimal disturbance of thin clouds and surface albedos to the aerosols detected by CALIPSO during the monsoon season (Redemann et al., 2012). Additionally, the robust and quick increase in aerosols in the South Asia has been verified by many previous studies (Suresh et al., 2013; Reddy et al., 2016; Srivastava, 2016).

Figure 2. Temporal variations in spatial average AOD from (a) CALIPSO and MODIS during the monsoon season, as well as (b) MODIS during pre-monsoon, monsoon, and dry seasons in the South Asian subcontinent from 2006 to 2015, respectively. The dashed lines indicate the linear fit line of the solid line with the same color; $p$ is the significance level; $k$ is the slope of the line trend of each time series.

As shown in Figure 3a, the average SSR (red line) in all-sky condition from CERES increased by 16.2 W m$^{-2}$ increments during the monsoon season in the South Asian subcontinent from 2006 to 2015. Moreover, the change in OSR (by
13.0 W m$^{-2}$) showed a highly negative correlation with that in the SSR during the study period (Figure 3b). These phenomena indicate that the attenuating effect of the atmosphere on solar radiation is gradually weakened, thereby increasing the incident solar radiation at the surface (brightening). There is the fact that the SSR is highly negatively correlated with aerosol loading (Folini and Wild, 2011). Therefore, the increase in AOD did not contribute to the recent brightening during the monsoon season in the South Asian subcontinent from 2006 to 2015.

Further, Figure 3a shows no distinct change in the SSR during the pre-monsoon season, even the reverse trend of SSR during the dry season (dimming) comparing to that during the monsoon season. The seasonal differences in OSR variations are similar to those of SSR (Figure 3b). Moreover, the trend of SSR during the dry season presented a higher significance level ($p<0.01$) than that during the monsoon season ($p=0.40$). Many reasons can contribute to the significance levels during different seasons. The average cloud fractions during the pre-monsoon and dry seasons are both approximately 20%, and much lower than that during the monsoon season (53.4%) according to CloudSat observations. Moreover, under clear-sky condition, aerosols control the change in the solar radiation that reaches the surface (Nair et al., 2016; Soni et al., 2016). Therefore, aerosols have greater weighted impact on solar radiation during the dry season than during the monsoon season. In addition, the resultant effects of aerosol, cloud, and water vapor are considerably complicated and varied due to the intensive atmospheric activities during the monsoon season. The pre-monsoon season is the transition season from the dry season to the monsoon season, and associated with the establishment of the southwestern wind regime over the South Asian subcontinent (Das et al., 2015). The atmospheric condition and radiation variations are not distinct comparing to the trend of SSR between the dry and monsoon seasons.
Figure 3. Temporal variations in spatial average (a) SSR and (b) OSR from CERES during pre-monsoon, monsoon, and dry seasons in the South Asian subcontinent from 2006 to 2015, respectively. The dashed lines indicate the linear fit line of the solid line with the same color; \( p \) is the significance level; \( k \) is the slope of the line trend of each time series.

5. Averaging region. I assume that line plots, as e.g. in Figure 2 or in Figure 3a, are averages taken over the region shown in the maps, e.g. in Figure 3c. If so, the average includes a substantial amount of sea and of the Himalaya. So the averages shown are not actually averages over India. The authors should comment on how this affects their results.

RE: Thank you for your suggestion. We defined the study domain within the South Asian subcontinent to restrict the effect of the Tibet Plateau and sea pixels on the results of this study, as shown in Figure 1. We have added the terrain map and showed the study region of this study in Figure 1 as the following:
Figure 1. Terrain map during the monsoon season in the South Asia and surrounding regions. The blue line indicates rivers; the study region is within the red line.

6. Statistical significance. While the authors give p-values in some of the figures, they do not address them in the text. Given that many p-values exceed 0.05 or even 0.1 the authors should comment on them. Also, there is no statistical significance given for any of the maps.

RE: Thank you for your suggestion. We have address the p-values in corresponding location in the text, and added comments about the significant level of relevant results. For example, the comment about shortwave surface radiation have been added at Paragraph 3 of Section 3.1 as the following:

Figure 2b shows the consistent increase in AOD during the pre-monsoon, monsoon, and dry seasons, respectively. The increment of AOD during the pre-monsoon and dry seasons (by 0.08 and 0.11, respectively) were greater than those during the monsoon season according to the MODIS observations from 2006 to 2015. The significant level of AOD variation detected by MODIS was higher during non-monsoon season ($p<0.01$) than during the monsoon season, which mainly due to the disturbance of widespread clouds during the monsoon season (Remer et al., 2005). The change in the average AOD detected by
CALIPSO showed a higher significance level ($p=0.07$) than that detected by MODIS due to the minimal disturbance of thin clouds and surface albedos to the aerosols detected by CALIPSO during the monsoon season (Redemann et al., 2012). Additionally, the robust and quick increase in aerosols in the South Asia has been verified by many previous studies (Suresh et al., 2013; Reddy et al., 2016; Srivastava, 2016).

The aerosol, cloud and water vapor variations is correlated with regional sources and meteorology. Consequently, solar dimming and brightening may be of local or regional nature, and is unavoidably inhomogeneous in space (Soni et al., 2016). Moreover, this study mainly focus on the total change of solar radiation at the surface and its significant reasons, not the relevant regional distribution. The recent transition from dimming to brightening is unstable and fluctuant in the South Asian subcontinent. However, due the crucial role of incident solar radiation on surface energy balance, any change with regional and global scale is worthy of our deep concern. We have added the comment of the solar radiation variations at the surface in space at Paragraph 3 of Section 4 as the following:

Additionally, the aerosol, cloud, and water vapor variations are correlated with regional sources and meteorology. Consequently, solar dimming and brightening may be of local or regional nature, and they are unavoidably inhomogeneous and unstable in space (Soni et al., 2016).

**Minor comments:**

1. p.6, l.6: What do you mean by "... with the scene that the lidar penetrates the entire atmosphere to exclude the effect of un-penetrated profile for lidar. "?

   **RE:** Thank you for your caution. Aerosol is mostly located at near-surface and the base of cloud. However, it is difficulty for CALIPSO lidar to penetrate the optically thick layer of optical depth exceeding 3-5 (Hu et al., 2009). In this case, CALIPSO many miss aerosol at near-surface due to the entirely attenuation of lidar signal. These case possible generate that the calculated AOD from CALISPO is less than the actual value. Therefore, we remove the unpenetrated profile for lidar before the calculation of AOD. We have added the corresponding explanation at Paragraph 1 Section 2.2 as the following:

   CALIPSO is highly sensitive to thin cirrus clouds, but cannot easily penetrate optically thick layers with optical depths exceeding 3–5 (Hu et al., 2009). In this case, CALIPSO determines no or low aerosol optical depth (AOD) at the
near-surface due to the complete attenuation of lidar signals. To exclude the effects of unpenetrated profiles for lidar, we calculate the average AOD from CALIPSO only on the basis of the scene in which the lidar penetrates the entire atmosphere.

2. p.6, l.10: What do you mean by "The low significant level occur in the line trend of AOD detected by MODIS, and mainly result from the disturbance of widespread cloud during monsoon season."?

RE: Thank you for your comment. The quick increase of aerosol in the South Asian subcontinent have been verified by many previous studies (Suresh et al., 2013; Reddy et al., 2016; Srivastava, 2016). The annual average cloud fraction during monsoon season is 53.4%, is more than twice of that during non-monsoon (about 20%) based no CloudSat observations. However, due to the impact of widespread cloud during monsoon season, the AOD variations show the lower significance level than that during non-monsoon, especially low broken cloud and thin cirrus (Mao et al., 2015). We have revised the corresponding description at Paragraph 1 of Section 3.1 as the following:

The significant level of AOD variation detected by MODIS was higher during non-monsoon season ($p<0.01$) than during the monsoon season, which mainly due to the disturbance of widespread clouds during the monsoon season (Remer et al., 2005).

3. p.6, l.12: What do you mean by "Moreover, aerosols mostly consist of dust, smoke, and polluted dust? These aerosols all present distinct absorbing effects on solar radiation, thereby attenuating the incoming solar radiation more intensively than do other types of aerosols in the surface. "What other types of aerosols? And why should aerosol mostly consist of dust, smoke (black carbon?), and polluted dust?"

RE: Thank you for your suggestion. CALIPSO classified the aerosols into six sub-types with high accurate, including clean marine, dust, polluted continental, clean continental, polluted dust and smoke. We have revised the corresponding description at Paragraph 1 of Section 3.1 as the following:

Moreover, CALIPSO can identify aerosol and cloud layer with ~90% accuracy. The sum of dust, smoke, and polluted dust aerosol loadings is close to the total aerosol loading (orange line in Figure 2a). This phenomenon indicates that aerosols mostly consist of dust, smoke, and polluted dust during the monsoon
season in the South Asian subcontinent, which is consistent with the previous study (Das et al., 2015).

4. p.7, l.21: "...the change of LW vertical heating rate is insignificant and similar to that in SW." However, for SW you claimed that it is significant. How then can it be similar?

**RE:** Thank you for your suggestion. I am sorry for the ambiguity. We have revised the corresponding description at Paragraph 4 of Section 3.2 as the following:

However, the change in LW vertical heating rate was not distinct compared with that of SW.

5. p.7, l.23: "...but cooling by 0.1 - 0.2 K d\(^{-1}\) within the cloud..." Looking at Figure 4, cooling seems rather on the order of 1 - 2 K d\(^{-1}\).

**RE:** Thank you for your suggestion. We have revised the corresponding description at Paragraph 4 of Section 3.2 as the following:

Correspondingly, the LW radiative heating increased by less than 0.2 K d\(^{-1}\) above 10 km, whereas that of heating decreased by 0.1–0.2 K d\(^{-1}\) within the cloud, especially below 4 km, from 2006 to 2015.

6. p.7, l.24: What do you mean by "In general, the net vertical heating rate weakens consistently, indicating that the changes of clouds physical characteristics shift the vertical heating and cooling, and weakening the vertical radiative effect of clouds. "?

**RE:** Thank you for your suggestion. We have revised the corresponding description at Paragraph 4 of Section 3.2 as the following:

In general, the net vertical heating rate weakened consistently, indicating that the changes in the clouds physical characteristics weakened their vertical radiative effect of clouds.

7. p.8, l.1: "particle number concentration", do you mean "cloud droplet number concentration"?

**RE:** Thank you for your suggestion. You are right. The "particle number concentration" have been revised to "cloud droplet number concentration"

8. Section 3.3: How is clear sky defined? I assume via 'few clouds in the pixel’, but a) what is few and b) how many pixels remain / what is the statistics?
RE: Thank you for your suggestion. We have defined the clear-sky condition at Paragraph 2 of Section 2.5 as the following:

We define the cloudy-sk condition for all products as the scenes in which the cloud mask from the CloudSat footprint exceeds 20, and the others are defined as clear-sky condition. Under this screened method of cloud, the probability of a false detection to CloudSat is less than 5% in comparison with that by CALIPSO (Mace and Zhang, 2014).

9. p.8, l.22: "Notably, high concentration of water vapor is favorable to generate more or stronger clouds and thus less radiation in surface and vice versa." One may debate this statement as clouds depend on two parameters: water vapor and temperature. The latter changes as well in your data, I assume, as SSR and heating rates change.

RE: Thank you for your suggestion. With the increase of SSR, the atmospheric temperature generally increased consistently. Similarly, high concentration of water vapor is favorable to generate more or stronger clouds comparing to the low concentration of water vapor. However, the resultant influence of atmospheric condition variations on solar radiation is complicated. In this study, we mainly focus on the impact of water vapor variation on solar radiation. We have revised the corresponding description at Paragraph 2 of Section 3.3 as the following:

Generally, aside from the effect of temperature, a high concentration of water vapor favors the generation of more or stronger convective clouds and results in decreased radiation at the surface, and vice versa (Yang et al., 2012).

10. p.9, l.14: "However, the conclusion of this study is significant even when overestimated factors are considered." Why should this be so?

RE: Thank you for your suggestion. We have revised the corresponding description at Paragraph 3 of Section 4 as the following:

In sum, the conclusions of this study are beneficial in the evaluation of the recent variations in aerosol, cloud, and water vapor and their resultant influences on solar radiation.

11. Figure 1a: Where does the line for "Dust, Smoke, Polluted Dust" come from? Which satellite / model data?
**RE:** Thank you for your suggestion. CALIPSO can identify the sub-types of aerosols, including clean marine, dust, polluted continental, clean continental, polluted dust and smoke. The sum of dust, smoke and polluted dust aerosol loading is calculated by CALIPSO data. We have revised the corresponding description at Paragraph 1 of Section 2.2 as the following:

Moreover, aerosols detected by CALIOP can be classified into the following six sub-types with high accuracy: clean marine, dust, polluted continental, clean continental, polluted dust, and smoke (Omar et al., 2009).

12. Figure 1b: y-axes should read "Radiative Fluxes" (not Fluxs).

**RE:** Thank you for your suggestion. We have revised "Fluxs" to "Fluxes".

13. Figure 7a: As for Figure 1b, the values are way too high to be monthly means. Colors in the figure and in the figure caption do not match.

**RE:** Thank you for your suggestion. The explanation about instantaneous and daily average radiation fluxes is provided in Question 1. Moreover, we have matched the colors in the figure and in the figure caption.
References


Soni, V. K., Pandithurai, G., and Pai, D. S.: Is there a transition of solar radiation from dimming to

Srivastava, R.: Trends in aerosol optical properties over South Asia, International Journal of Climatology, 37, n/a-n/a, 2016.


Estimating the effects of aerosol, cloud, and water vapor on the recent brightening during the South Asian monsoon season

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Abstract. South Asia is experiencing a leveling-off trend of solar radiation and even a transition from dimming to brightening. As the only significant energy source for the global ecosystem, any change in incident solar radiation affects our habitats profoundly. This process is significantly complicated because of the active atmospheric action during the monsoon season. Here, we use observations from multiple sensors in the A-Train satellite constellation to evaluate the effects of aerosol, cloud, and water vapor variations on the recent changes in surface solar radiation during the monsoon season (June–September) in the South Asian subcontinent from 2006 to 2015. Results show that during this period, the surface shortwave radiation (SSR) and outgoing shortwave (SW) radiation increased by 16.2 W m\(^{-2}\) and decreased by 13.0 W m\(^{-2}\), respectively. However, the increase in spatial average aerosol optical depth is inconsistent with the variation in SSR in the South Asian subcontinent. Instead, the decreases in the amount of water vapor and clouds significantly contributed to brightening, thus further affecting the surface warming in the South Asian subcontinent. Clouds generally reduced and thinned by approximately 9.4% and 182 m, respectively, with the decrease in cloud water path (by 53.4 g m\(^{-2}\)) and particle number concentration under cloud-sky condition. Given the change in the clouds, the atmospheric vertical SW heat rating decreased by 0.3–0.4 K d\(^{-1}\) at an altitude of 5–15 km, whereas the SW heat rating slightly increased at the low-cloud and near-surface areas (below 4 km) from 2006 to 2015. The SW cloud radiative effect decreased by approximately 45.5 W m\(^{-2}\) at the surface. Moreover, the precipitable water under clear-sky condition decreased by approximately 2.8 mm over the brightening period. Correspondingly, solar brightening increased by roughly 2.5 W m\(^{-2}\) owing to the weakened absorption. Overall, the decreases in water vapor and clouds resulted in the increased absorption of direct solar radiation at the surface and subsequent surface brightening. Hence, brightening may play a prominent role in modulating the warming and rainfall variations in the South Asian subcontinent.
1 Transition from Dimming to Brightening in South Asia

Solar radiation incidence on the Earth’s surface plays a fundamental determinant role in climate and life on our planet (IPCC, 2013). Surface solar radiation is a major component of the surface energy balance and governs many diverse surface processes, such as evaporation and associated hydrological components, snow and glacier melting, plant photosynthesis and related terrestrial carbon uptake, as well as the diurnal and seasonal courses of surface temperatures. Negative trends in the downwelling surface solar radiation are collectively called “dimming,” whereas positive trends are called “brightening” (Wild et al., 2005). Any change in the amount of solar radiation profoundly affects the temperature field, atmospheric and oceanic general circulation, and hydrological cycle (Haywood et al., 2011).

Widespread reduction in the annual average surface solar radiation from the 1960s to the 1980s has been reported by many researchers at the global and regional scales, including those from America, Europe, and China (Liepert, 2002; Wild et al., 2005). Subsequently, the term “brightening” was coined to emphasize that global solar radiation no longer declines at many sites after the late 1980s (Wild et al., 2005). Long et al. (2009) found that solar dimming reversed at an increasing trend of 6 W m⁻² per decade in the continental United States in 1995–2007. Wild (2009) presented that the globally averaged trends in the 1980s typically reversed from dimming to brightening; the study reported trends of 2.2–6.6 W m⁻² per decade from the 1980s to the 2000s. However, recent studies indicate that the developments in dimming and brightening after 2000 show mixed tendencies. Wild et al. (2009) reported a continuation of brightening at sites in Europe, United States, and parts of Asia, a leveling-off at sites in Japan and Antarctica, and indications of renewed dimming in China. Conversely, the most recent related study shows that brightening has been continuing in China after 2000 (Wang and Wild, 2016).

South Asia is endowed with abundant solar energy because of its geographic position in the tropical belt (Soni et al., 2016). Dimming or brightening in South Asia is more evident and complicated than that in other regions, and its effects on regional and global climate and ecosystem are amplified by the monsoon circulation in the country (Padma et al., 2007; Wild, 2012). The South Asian monsoon occurs between June and September, which correspond to the monsoon onset and end times, respectively (Turner and Annamalai, 2012). Contrary to the variable trend in surface solar radiation in other regions globally, Padma et al. (2007) found that continued dimming of −8.6 W m⁻² per decade occurred in India from 1981 to 2004 according to 12 stations over the Indian region. Wild et al. (2009) proposed to focus on the slight tendency toward the stabilization of surface solar radiation since the late 1990s. Furthermore, Soni et al. (2016) observed a trend reversal and partial recovery from dimming to brightening in India around 2001.

Various mechanisms can potentially contribute to dimming and brightening. Changes in surface solar radiation can be caused by either external changes in the amount of solar radiation incident on the planet at the top of atmosphere (TOA) or internal changes (within the climate system) in the transparency of the atmosphere, which modifies the solar beam on its way to the Earth’s surface. However, Willson and Mordvinov (2003) verified that the decadal dimming or brightening cannot be explained by changes in the luminosity of the sun because these changes are at least an order of magnitude smaller than
those in climate system. Therefore, the observed dimming or brightening has to originate from alterations in the transparency of the atmosphere, which depends on the presence of clouds, aerosols, and radiatively active gases, particularly water vapor, which is a strong absorber of solar radiation (Kim and Ramanathan, 2008).

Kvalevåg and Myhre (2007) concluded that the major contributor to dimming is aerosols (~2.4 W m⁻²) and that the secondary effect is the increase in gas concentrations (~0.64 W m⁻²), including those of tropospheric ozone and water vapor, since pre-industrial times; they also identified NO₂, CH₄, and CO₂ as minor contributors. Many studies have proposed that the changes in atmospheric aerosol loading, cloud cover, and cloud properties are the main factors that determine solar dimming and brightening (Wild, 2009; Soni et al., 2016). Liepert (2002) showed that the decrease in global radiation in 1961–1990 is attributable to the increases in cloud optical thickness and the direct effects of aerosols, which reduced solar radiation by 18 and 8 W m⁻², respectively. Kambezidis et al. (2012) argued that the decline in surface solar radiation in South Asia is attributable to the increase in the amount of anthropogenic aerosols during the last 30 years of the 20th century. This deduction is supported by the agreement between the observed decadal changes in anthropogenic aerosol emissions and trends in global solar radiation (Wild, 2009; Folini and Wild, 2011). Moreover, solar dimming and brightening may be of local or regional nature and is unavoidably influenced by regional sources and meteorology (Soni et al., 2016). Here, we focus mainly on the dimming and brightening caused by changes in aerosols, water vapor, and clouds, which are the dominant factors that alter atmospheric transparency, and thereby regulate the solar radiation incident on the Earth’s surface (Wild, 2009).

In this study, we use observations from multiple sensors in the A-Train satellite constellation to evaluate the effects of aerosol, cloud, and water vapor variations on the recent changes in solar radiation at the surface during the monsoon season (June–September) in the South Asian subcontinent from 2006 to 2015. This study mainly aims to identify the possible reasons for the changes in dimming and brightening and assess the connection between the variations in aerosols, clouds, and water vapor and the recent brightening in the South Asian subcontinent.

2 Data and Methods

2.1 CloudSat data

Cloud Profile Radar (CPR) is an active millimeter-wave radar and is the only instrument on CloudSat, which was launched into the A-Train constellation in April 2006 (Mace et al., 2009). CPR has a 1.3 km cross-track and a 1.7 km along-track footprint resolution, and its effective vertical resolution at nadir is 240 m (Pan et al., 2017). CloudSat CPR is well suited for sensing a wide variety of cloud systems, from cirrus and stratus to deep convective systems, and shows slight sensitivity to the time of day or season (Rajeevan et al., 2012). However, the estimated operational sensitivity of CloudSat CPR (~32 dBZ to ~30 dBZ) is insufficient for observing thin cirrus clouds with low ice water content (IWC) (Mace et al., 2009). CloudSat can infer the cloud microphysical characteristics on the basis of the backscatter return signals measured by
CloudSat CPR, including cloud particle number concentration, size, shape, and phase (Heymsfield et al., 2010). Further, by combining with a broadband radiative flux model known as BUGSRad, CloudSat can be used to quantify the three-dimensional (3D) information of cloud macrophysical and microphysical characteristics, as well as the corresponding cloud radiative effect (CRE) (L'Ecuyer et al., 2008).

The CloudSat Radar-Only Cloud Water Content (2B-CWC-RO) product contains retrieved estimates of cloud liquid water content (LWC) and IWC, cloud liquid number concentration (LNC) and ice particle number concentration (INC), cloud liquid effective radius (LER) and ice particle effective radius (IER), and related quantities for each radar profile measured by CPR on CloudSat (Woods et al., 2008). Considering the insensitivity of CloudSat to thin cirrus, Austin et al. (2009) showed that 2B-CWC-RO produces typical biases in IWC of −40% to +25% versus the observationally derived synthetic data. However, L'Ecuyer et al. (2008) proposed that the impact of the thin cirrus that is not detected by CloudSat on shortwave (SW) radiation at the surface is −1.2 W m$^{-2}$ in a globally averaged experiment on radiative fluxes; this value is much smaller than the impact value of low clouds. Moreover, Christensen et al. (2013) found that CloudSat-derived cloud liquid water path (LWP) is generally too high and systematically exceed those from Moderate Resolution Imaging Spectroradiometer (MODIS) by approximately 50%. Most previous studies have verified the CloudSat 2B-CWC-RO product, thereby allowing us to discuss the 3D microphysical characteristics of clouds in detail (Austin et al., 2009; Rajeevan et al., 2012). Therefore, the products containing 3D cloud characteristics from CloudSat are significant in evaluating cloud characteristics and their variations.

Furthermore, retrieved profiles of cloud microphysical properties form the basis of the algorithm of another data product (2B-FLXHR), which consists of high vertical resolution profiles of radiative fluxes and atmospheric heating rates. L'Ecuyer et al. (2008) detected biases between the radiative data of 2B-FLXHR from CloudSat and those from Clouds and Earth’s Radiant Energy System (CERES) with monthly 5° means in the global scale. The biases of outgoing shortwave radiation (OSR), outgoing longwave (LW) radiation, surface shortwave radiation (SSR), and surface LW radiation are less than 0.1, 5.5, 13, and 16 W m$^{-2}$, respectively. Fortunately, the uncertainties in 2B-FLXHR fluxes decrease significantly for long time scale averages (L'Ecuyer et al., 2008). Therefore, the CRE derived from CloudSat can credibly describe the radiative effect of clouds, especially in large space and time scales.

### 2.2 CALIPSO data

Cloud-Aerosol Lidar and Infrared Pathfinder Satellite Observations (CALIPSO) was also launched in April 2006 (Winker et al., 2007). CALIPSO mainly loads Cloud Aerosol Lidar with Orthogonal Polarization (CALIOP), which is an excellent active two-wavelength (532 and 1064 nm) polarization lidar (Winker et al., 2007). The layers detected by CALIOP are correctly identified with a high degree of confidence (> 90%) by cloud–aerosol mask analysis (Liu et al., 2009). Moreover, aerosols detected by CALIOP can be classified into the following six sub-types with high accuracy: clean marine, dust, polluted continental, clean continental, polluted dust, and smoke (Omar et al., 2009). Therefore, CALIPSO provides 3D...
scientific materials for and perspective on studying the changes, interactions, and transportation of aerosols and clouds at a global scale (Winker et al., 2010; Pan et al., 2015). In this study, CALIPSO Level 2 aerosol layer products are used to describe the recent changes in aerosol loading and its mostly sub-type component in the South Asian subcontinent. Moreover, the observations are compared with those of MODIS. CALIPSO is highly sensitive to thin cirrus clouds, but cannot easily penetrate optically thick layers with optical depths exceeding 3–5 (Hu et al., 2009). In this case, CALIPSO determines no or low aerosol optical depth (AOD) at the near-surface due to the complete attenuation of lidar signals. To exclude the effects of unpenetrated profiles for lidar, we calculate the average AOD from CALIPSO only on the basis of the scene in which the lidar penetrates the entire atmosphere.

2.3 CERES/MODIS data

Aqua became a member of the A-Train constellation in May 2002, with the load of CERES, MODIS, and four other sensors (Remer et al., 2005; Wielicki et al., 2015). CERES/Aqua Edition 4A Single Scanner Footprint (SSF) and MODIS/Aqua Edition 6 Atmosphere Level 3 Joint Products (MYD08) data are used in this study. CERES SSF data sets combine CERES radiation measurements, cloud and aerosol microphysical property retrievals based on observations of MODIS, and ancillary meteorology fields to form a comprehensive, high-quality compilation of satellite-derived cloud, aerosol, and radiation budget information for radiation and climate studies (Loeb and Manalosmith, 2005). CERES reports the observed SW and LW solar radiances with precisions approaching 1.0% (Loeb and Manalosmith, 2005). Moreover, this study uses land–ocean AOD at 0.55 μm from MYD08. On the basis of its comparison with the Aerosol Robotic Network, the MODIS AOD has been verified with absolute and relative errors of 0.05 and 15%, respectively, mainly from the retrieval algorithm (Levy et al., 2013; Wang et al., 2017). AOD data have been widely used and verified by many studies because of their excellent capability of deriving AOD from dark to bright surfaces (Remer et al., 2005; Munchak et al., 2013). The most significant effects of cloud contamination on passive MODIS and CERES have to be emphasized. For example, artificially high AOD can be detected by MODIS due to the presence of thin clouds undetected by MODIS (Grandey et al., 2013). Moreover, the scattering of light by inhomogeneous clouds possibly generates uncertainty in the passive wide-field observations of MODIS and CERES, which are mainly caused by broken cloud systems (Grandey et al., 2013; Christensen et al., 2016). Therefore, the uncertainties of MODIS and CERES observations are more complicated in widespread-cloud conditions than in insufficient-cloud conditions.

2.4 Radiative transfer model

BUGSrad is the official radiative transfer model used by the product of CloudSat 2B-FLXHR (Fu and Liou, 1992; L’Ecuyer et al., 2008). This model is based on the two-stream doubling-adding solution to the radiative transfer equation with the assumption of a plane-parallel atmosphere (Stephens et al., 2001). BUGSrad computes molecular absorption and scattering properties on the basis of correlated-k formulation (Fu and Liou, 1992). The calculation of BUGSrad is parallelly applied over six SW bands, and a constant hemisphere formulation is applied to 12 LW bands. These bands are appropriately
weighted and combined into the two broadband flux estimates that are ultimately reported, one covering the SW at 0–4 µm and the other covering the LW above 4 µm. According to the comparison experiment with observations of the Atmospheric Remotely-Sensed Clouds Locations of the Atmospheric Radiation Measurement program, the mean biases of SW and LW in clear-sky are 1.2 and 2.2 W m⁻² at TOA, respectively (Stephens et al., 2001; Stephens et al., 2003).

### 2.5 Principle and methods

In this study, multiple sensors from A-Train are used to investigate the possible reasons for the dimming and brightening variations during the monsoon season (June–September) in the South Asian subcontinent from 2006 to 2015. We define the study domain within the South Asian subcontinent to restrict the effect of the surrounding land and sea pixels on the results of this study, as shown in Figure 1. CloudSat, CALIPSO, and Aqua in A-Train are maintained in a tight orbital coordination within 176 s (Stephens et al., 2017). This setting allows near-simultaneous observations of a wide variety of atmospheric and surface parameters, facilitating the comparison of satellite between each other, allowing for even more comprehensive studies of climate. The MYD08 and Level 2 aerosol layer products from MODIS and CALIPSO are used to describe the recent changes in aerosol loading in the South Asian subcontinent. The SSF product from CERES/Aqua is considered for evaluating the radiation energy variations. The CloudSat 2B-CLDCLASS, 2B-CWC-RO, and 2B-FLXHR products are used to quantify the 3D changes in cloud macrophysical and microphysical characteristics and corresponding variable radiative effects of clouds. Moreover, Environmental conditions are obtained from the European Center for Medium-range Weather Forecast-AUXiliary analysis (ECMWF-AUX) product (Uppala et al., 2005). The CRE inferred from CERES is always disturbed by water vapor, aerosol, and the limited capability of MODIS to detect thin cirrus; furthermore, CERES considers the impact of multi-layer clouds insufficiently (Sohn et al., 2010). Therefore, this study uses the CloudSat 2B-FLXHR to evaluate the effects of cloud variations on radiation.

This study focuses on the effect of aerosol, cloud, and water vapor variations during the monsoon season. Due to the increased convective activity, the changes in aerosol, cloud, and water vapor and their impact on solar radiation possibly occur more obviously during the monsoon season than during non-monsoon seasons (Turner and Annamalai, 2012). Additionally, the information about aerosol, cloud, and water vapor is calculated on the basis of instantaneous observations, which we then use to ensure consistent conditions for the observations of multiple satellites. The satellites or sensors of the A-Train cross the equator in the daytime at 13:30 local time, at which point the incoming solar radiation is maximum within the day (Liou, 2002). Therefore, the instantaneous observations of radiative fluxes used in this study are greater than the daily average. The results are more representative of the atmospheric condition variations at nominally 13:30 local time than the daily average. For the method of spatial and temporal averages, we firstly obtain the spatial distribution of temporal average within a 1° × 1° grid during the monsoon season. We then determine the spatial average on the basis of the above temporal average in the entire South Asian subcontinent. For the spatial patterns of temporal changes, we identify the results of the linear temporal changes in atmospheric conditions during the monsoon season from 2006 to 2015.
Strict selection procedures are implemented to control the quality of the data products and thereby ensure credible conclusions. The CPR cloud mask and radar reflectivity from 2B-GEOPROF are set to be more than 20 and −28 dBZ, respectively, and the quality flag from the 2B-CLDCLASS is identified as confidence to CloudSat (Mace and Zhang, 2014). We define the cloud-sky condition for all products as the scenes in which the cloud mask from the CloudSat footprint exceeds 20, and the others are defined as clear-sky condition. Under this screened method of cloud, the probability of a false detection to CloudSat is less than 5% in comparison with that by CALIPSO (Mace and Zhang, 2014). The all-sky condition is the total of cloud-sky and clear-sky condition. Data quality for CALIPSO is maintained by screening the cloud layer with a high degree of confidence (> 90%) (Hu et al., 2009). We only use the radiative data from CERES with a sensor viewing zenith angle less than 60° at the surface to restrict the uncertainty from the satellite non-nadir point (Christopher and Zhang, 2002). Due to the SW solar radiation (< 5 μm) comprising more than 99.5% of total solar energy, it is assumed as the total incoming solar radiation at the TOA and surface in this study. Although the BUGSrad defines the SW from 0–4 μm, the solar energy at 4–5 μm is excessively low (approximately 0.39% of total solar energy) (Liou, 2002). Therefore, the effect of this difference is ignorable between the definitions of SW between CERES and BUGSrad. Surface solar radiation is analyzed in this study in a relative sense with variable aerosols, clouds, and water vapor; thus, this study focuses mostly on the relative changes in solar radiation but less on the absolute values of solar radiation.

CRE has been widely used to quantify the degree of cloud–radiation interactions (Henderson et al., 2013). CRE is the net (down minus up) flux difference between the all-sky condition and clear-sky condition on the atmosphere, surface, or TOA, as shown in Eq. (1):

\[
CRE = (F^\downarrow - F^\uparrow)_{\text{All-Sky}} - (F^\downarrow - F^\uparrow)_{\text{Clear-Sky}},
\]  

(1)

where \( F^\downarrow \) and \( F^\uparrow \) are the downwelling and upwelling radiative fluxes at the TOA and surface, respectively. One aim of this study is to quantify the impact of cloud variations on TOA and surface fluxes. Therefore, radiative data from the CloudSat 2B-FLXHR products at the surface and TOA are used. Moreover, this study uses precipitable water (PW) variation to describe the column changes of water vapor in the atmosphere. PW is the total water vapor in the atmospheric column, which is calculated on the basis of the data on environmental parameters from the ECMWF-AUX product, as shown in Eq. (2):

\[
PW = \frac{1}{g_0} \int_{P_{\text{TOA}}}^{P_{\text{Surface}}} q(P) dP,
\]  

(2)

where \( g_0 = 9.80665 \text{ m s}^{-2} \) is the standard acceleration due to gravity at the mean sea level (m.s.l.), \( P \) is the atmospheric pressure, and \( q(P) \) is the specific humidity of air as a function of atmospheric pressure (Bock et al., 2010). The PW data from ECMWF-AUX have been verified with a bias of approximately −1 mm depending on water vapor amount by comparing observations from 21 ground-based Global Positioning System receiving stations and 14 radiosonde stations (Bock et al.,...
Therefore, although the PW inferred from ECMWF-AUX is generally slightly lower than that under actual condition, it can adequately indicate the change in water vapor amount in the South Asian subcontinent.

We accurately evaluate the effect of water vapor variations on solar radiation using the official radiative transfer model of CloudSat 2B-FLXHR (BUGSrad). By using BUGSrad, it is convenient to compare the results of the radiative transfer model with those of the CloudSat product. We use the environmental parameters provided by CloudSat products as the input of the BUGSrad, including temperature, pressure, specific humidity and cloud mask, and so on. Then, we only change the average water vapor amount during the monsoon season, and other environmental parameters are set as the inter-annual average values during the monsoon season from 2006 to 2015 to exclude the effect of other factors. For the radiative transfer model, we directly use the averaged environmental parameters during the monsoon season to evaluate the radiative effect of water vapor by considering the quantity of the calculation. Due to the difference of averaged methods which were used by CloudSat and BUGSrad to evaluate the radiative effect of water vapor, there may be some difference between the radiative fluxes calculated based on CloudSat and BUGSrad.

3 Results and Discussion

3.1 Changes in aerosols and solar radiation

Figure 2 shows the recent changes in aerosol loading during different seasons, namely, pre-monsoon, monsoon, and dry seasons, respectively. These three seasons occur in the periods of March–May, June–September, and October–February, respectively. The aerosol loading increased, and the change in the average AOD detected by CALIPSO (approximately 0.13) was more intense than that detected by MODIS (approximately 0.05) during the monsoon season in the South Asian subcontinent from 2006 to 2015 (Figure 2a). These phenomena are mainly due to the effect of aerosols under clouds, which can be detected by CALIPSO but missed by MODIS because of MODIS only inferring the aerosol loading under clear-sky condition (Winker et al., 2010). Moreover, CALIPSO can identify aerosol and cloud layer with ~90% accuracy. The sum of dust, smoke, and polluted dust aerosol loadings is close to the total aerosol loading (orange line in Figure 2a). This phenomenon indicates that aerosols mostly consist of dust, smoke, and polluted dust during the monsoon season in the South Asian subcontinent, which is consistent with the previous study (Das et al., 2015). These aerosols all present distinct absorbing effects on solar radiation, thereby attenuating the incoming solar radiation more intensively than other types of aerosols (Logan et al., 2013).

Figure 2b shows the consistent increase in AOD during the pre-monsoon, monsoon, and dry seasons, respectively. The increment of AOD during the pre-monsoon and dry seasons (by 0.08 and 0.11, respectively) were greater than those during the monsoon season according to the MODIS observations from 2006 to 2015. The significant level of AOD variation detected by MODIS was higher during non-monsoon season \(p<0.01\) than during the monsoon season, which mainly due to the disturbance of widespread clouds during the monsoon season (Remer et al., 2005). The change in the average AOD
detected by CALIPSO showed a higher significance level ($p=0.07$) than that detected by MODIS due to the minimal disturbance of thin clouds and surface albedos to the aerosols detected by CALIPSO during the monsoon season (Redemann et al., 2012). Additionally, the robust and quick increase in aerosols in the South Asia has been verified by many previous studies (Suresh et al., 2013; Reddy et al., 2016; Srivastava, 2016).

As shown in Figure 3a, the average SSR (red line) in all-sky condition from CERES increased by 16.2 W m$^{-2}$ increments during the monsoon season in the South Asian subcontinent from 2006 to 2015. Moreover, the change in OSR (by 13.0 W m$^{-2}$) showed a highly negative correlation with that in the SSR during the study period (Figure 3b). These phenomena indicate that the attenuating effect of the atmosphere on solar radiation is gradually weakened, thereby increasing the incident solar radiation at the surface (brightening). There is the fact that the SSR is highly negatively correlated with aerosol loading (Folini and Wild, 2011). Therefore, the increase in AOD did not contribute to the recent brightening during the monsoon season in the South Asian subcontinent from 2006 to 2015.

Further, Figure 3a shows no distinct change in the SSR during the pre-monsoon season, even the reverse trend of SSR during the dry season (dimming) comparing to that during the monsoon season. The seasonal differences in OSR variations are similar to those of SSR (Figure 3b). Moreover, the trend of SSR during the dry season presented a higher significance level ($p<0.01$) than that during the monsoon season ($p=0.40$). Many reasons can contribute to the significance levels during different seasons. The average cloud fractions during the pre-monsoon and dry seasons are both approximately 20%, and much lower than that during the monsoon season (53.4%) according to CloudSat observations. Moreover, under clear-sky condition, aerosols control the change in the solar radiation that reaches the surface (Nair et al., 2016; Soni et al., 2016). Therefore, aerosols have greater weighted impact on solar radiation during the dry season than during the monsoon season. In addition, the resultant effects of aerosol, cloud, and water vapor are considerably complicated and varied due to the intensive atmospheric activities during the monsoon season. The pre-monsoon season is the transition season from the dry season to the monsoon season, and associated with the establishment of the southwestern wind regime over the South Asian subcontinent (Das et al., 2015). The atmospheric condition and radiation variations are not distinct comparing to the trend of SSR between the dry and monsoon seasons.

### 3.2 Effect of clouds

Figure 4 shows the inter-annual variations in the average cloud vertical frequency distribution in all-sky condition and the cloud microphysical parameters in cloud-sky condition during the monsoon season in the South Asian subcontinent from 2006 to 2015. The cloud vertical frequency sequentially decreased during the monsoon season in the South Asian subcontinent from 2006 to 2015, with a maximum decrement of 7.1%, which contributed to approximately 30% of the maximum cloud vertical frequency at approximately 12 km m.s.l. (Figure 4a). The cloud water content and number concentration of liquid and ice in cloud-sky condition showed a consistently decreasing trend, with maximum decrements of 18.0 mg m$^{-3}$ of LWC and 1.4 mg m$^{-3}$ of IWC, as well as approximately 6.0 cm$^{-3}$ of LNC and 2.0 L$^{-3}$ of INC (Figures 4d and
4g). These changes are consistent with those in the cloud vertical frequency during the study period. However, the LER and IER variations were not distinct and not highly consistent with the changes in the cloud vertical frequency (Figures 4c and 4f). Therefore, the clouds decreased with the reduced cloud water content and number concentration, but no clear change was observed in the particle effective radius during the monsoon season in the South Asian subcontinent from 2006 to 2015.

As shown in Figure 5a, the total cloud fraction detected by CloudSat declined by 8.8% ($p=0.07$) during the monsoon season in the South Asian subcontinent from 2006 to 2015. Furthermore, the change in CWP in cloud-sky condition was consistent with that of the total cloud fraction with a decrement of 54.0 g m$^{-2}$ ($p<0.01$). The uppermost cloud top height (CTH) declined by 661 m ($p=0.08$), and the cloud geometrical depth (CGD) declined by 280 m ($p=0.21$) with partial fluctuations in 2009 and 2010. By contrast, an insignificant trend occurred in the lowermost cloud base height (CBH) variation. We then further analyze the spatial variations in cloud fraction, cloud height (CH), CWP, and CGD during the monsoon season in the South Asian subcontinent from 2006 to 2015. Figures 5c–5f illustrate the consistent decreases in cloud fraction, CH, CWP, and CGD, especially for the cloud fraction and CWP with significant and consistent changes in space. CH and CGD increased in certain parts of the western coastal area in the South Asian subcontinent. In general, we can conclude that the clouds decreased and thinned with decreases in water content and particle number concentration during the monsoon season in the South Asian subcontinent from 2006 to 2015, thereby weakening the regulation of cloud on radiation.

The factors contributing to the changes in clouds are complicated and varied in the South Asian subcontinent. Ackerman et al. (2000) verified that aerosols with intensive absorption may lead to evaporation in the cloud layers (the aerosol semi-direct effect), burning off the clouds in the South Asia, including widespread dust, polluted dust, and smoke. Bollasina et al. (2011) showed the anthropogenic aerosol emissions weakened the South Asian summer monsoon, causing a decrease in the observed occurrence and amount of precipitation. In the global scale, observed and simulated cloud change patterns are consistent with the poleward retreat of mid-latitude storm tracks, which is attributed to the increasing greenhouse gas concentrations and recovery from volcanic radiative cooling (Norris et al., 2016). Actual, with the reduction of clouds, the recent weakened rainfall and dried monsoon during the summer monsoon season have been verified by many previous studies (Bollasina et al., 2011; Turner and Annamalai, 2012; Annamalai et al., 2013). However, the decreasing clouds also cause the increase in SSR, possibly enhancing the generation of rainfall (Wild, 2009). Therefore, the weakened rainfall was not steady, and can be attributed to the resultant force of many factors, including the changes in monsoon intensity, intrinsic cloud properties, and anthropogenic aerosols, to name a few.

Consequently, we investigate the radiative force of cloud vertical variations to determine their possible contribution to the recent rapid brightening (Figure 6). We use the vertical heating rates in all-sky condition to indicate the change in the vertically radiative effects of clouds (Fu and Liou, 1992). Given the changes in cloud macrophysical and microphysical characteristics, the SW heat rating decreased by 0.3–0.4 K d$^{-1}$ at an altitude of 5–15 km from 2006 to 2015, whereas a few increase in SW heat rating occurred at the low-cloud and near-surface areas (below 4 km). With the reduction of clouds, the total cloud reflection and absorption (mainly located at 5–15 km) weakened consistently, thus enhancing atmospheric
transmission, causing more SW radiation than before to reach the near-surface area (Henderson et al., 2013). Due to the extremely little cloud detected by CloudSat above 15 km, no significant change in heat rating occurred above 15 km, as shown in the cloud vertical frequency distribution in Figure 4a. However, the change in LW vertical heating rate was not distinct compared with that of SW. Correspondingly, the LW radiative heating increased by less than 0.2 K d\(^{-1}\) above 10 km, whereas that of heating decreased by 0.1–0.2 K d\(^{-1}\) within the cloud, especially below 4 km, from 2006 to 2015. The total cloud LW absorption emitted from the surface decreased with the reduction of clouds, thus leading to a weakened LW heating rate within the cloud layer (mainly below 10 km), and more LW emission on the top cloud layer above 10 km (L'Ecuyer et al., 2008). In general, the net vertical heating rate weakened consistently, indicating that the changes in the clouds physical characteristics weakened their vertical radiative effect of clouds.

As shown in Figure 7a, the SW CRE at the surface and TOA weakened by approximately 42.1 and 38.5 W m\(^{-2}\) with significance levels \(p\) of 0.05 and 0.04, respectively. Solar radiation consequently increased and reached at the surface during the monsoon season in the South Asian subcontinent from 2006 to 2015. The decrease in SW CRE at the surface and TOA was highly consistent with that of the total cloud faction (red line in Figure 7a). Although the CRE derived from CloudSat largely ignores the contribution of high thin clouds, this contribution is much smaller than that of low clouds (L'Ecuyer et al., 2008; Henderson et al., 2013). Moreover, SW CRE (Figure 7b) at the surface is negatively correlated with SSR (Figure 7c). This observation is attributed to the spatial variations in the radiative effect at the surface caused by cloud variables. In general, clouds were reduced and thinned by approximately 8.8% and 280 m with the decrease in cloud water content by 54.0 g m\(^{-2}\) and particle number concentration in cloud-sky condition. Consequently, the SW radiative effect of the clouds decreased by approximately 42.1 W m\(^{-2}\), and the absorption of direct solar radiation at the surface increased, thus leading to subsequent surface brightening.

### 3.3 Effect of water vapor

In this section, we evaluate the effect of water vapor on the recent brightening in the South Asian subcontinent. We focus mainly on the effect of water vapor on solar radiation in clear-sky condition to eliminate the disturbance of clouds in the evaluation of water vapor in all-sky condition. The vertical relative humidity (RH) in all-sky condition gradually decreased with a maximum decrement of approximately 10% during the monsoon season in the South Asian subcontinent from 2006 to 2015 (Figure 8a). Moreover, the change in RH in all-sky condition was highly consistent with that in clear-sky condition (Figures 8a and 8c). Figure 8b shows the spatial variation in PW in all-sky condition during the monsoon season in the South Asian subcontinent from 2006 to 2015; this spatial variation is highly consistent with that in clear-sky condition (Figure 8d). PW consistently decreased in space with a maximum decrement of 6.6 mm in the South Asian subcontinent, especially in the western and central regions. Although a small increase in water vapor (less than 2 mm) occurred in part of the South Asian region, its increment was much lower than the decrease in water vapor in most regions of the South Asian subcontinent. The decrease in atmospheric water vapor may be attributed to the weakened monsoon intensity in the recent
decade during the monsoon season in the South Asia (Turner and Annamalai, 2012).

Figure 9a shows that radiation is highly and negatively correlated with PW from ECMWF-AUX, thereby indicating that water vapor is a substantial controlling factor in solar radiation variability. This control lies in the direct and indirect effects. The former refers to the absorption of solar radiation, in which the PW decreased by nearly 2.8 mm ($p=0.13$) over the brightening period. According to the evaluation of BUGSrad, solar brightening was greater than 2.5 W m$^{-2}$ ($p=0.14$) because of the weakened absorption caused by the decrease in water vapor; the value was much lower than that of cloud variations (42.1 W m$^{-2}$). The SSR under clear-sky condition from CloudSat increased by approximately 4.8 W m$^{-2}$ ($p=0.05$). Except for the contribution of water vapor variations, this radiative change was possibly caused by the change in atmospheric conditions and incoming solar radiation provided by ECMWF and CloudSat (Yang et al., 2012). The spatial variation in SSR under clear-sky condition was highly negative and consistent with that of PW, especially in the western and central regions of the South Asian subcontinent (Figure 9b), which partly consist of deserts. Moreover, in all-sky condition, there are the indirect effects of water vapor, which is named as water vapor-cloud interaction. Generally, aside from the effect of temperature, a high concentration of water vapor favors the generation of more or stronger convective clouds and results in decreased radiation at the surface, and vice versa (Yang et al., 2012). Additionally, water vapor regulates cloud amount by affecting the aerosol-cloud interaction, which generally shows a positive relationship under moist atmospheric conditions (Chen et al., 2014). Therefore, the impact of water vapor in all-sky condition is more complicated than that in clear-sky condition owing to the direct and indirect effects of water vapor on solar radiation.

### 3.4 Climate and environmental implications

Variations in surface solar radiation can potentially significantly impact the climate system based on the long temporal correlation between solar radiation and climate factors, especially in temperature and rainfall. Observed dimming was suggested to be responsible for the absence of a significant temperature rise between the 1950s and the 1980s in various parts of the world, such as in the Arctic, China, America, and India (Ramanathan et al., 2005; Wild, 2009). However, the suppression of global warming over global land surfaces only lasted into the 1980s with the transition of dimming to brightening (Wild et al., 2005). This condition indicates that brightening might have significantly contributed to the recent rapid warming after the 1980s according to ground-based observations (Philipona et al., 2009). This temperature evolution satisfactorily fits the observational surface solar radiation variations, and points to the crucial role that dimming and brightening may play in determining global warming. Moreover, surface solar radiation variations induce changes in the surface net radiation, thereby altering the energy available for evaporation, which equals precipitation in the global annual mean. A decrease in evaporation and the same globally averaged reduction in precipitation with the globally averaged dimming occurred in the 1980s (Liepert, 2002). Wild et al. (2009) showed that the increase in available surface energy is quantitatively consistent with the observed substantial increase in land precipitation (3.5 mm y$^{-1}$ between 1986 and 2000) and the associated intensification of the land-based hydrological cycle.
The recent rapid warming in South Asia has been verified by many studies (Turner and Annamalai, 2012; Annamalai et al., 2013). Moreover, ECMWF-AUX shows that the 2 m temperature increased by 0.14 K y\(^{-1}\), even for surface temperature with the increment of 0.4 K y\(^{-1}\) during the monsoon season in the South Asian subcontinent from 2006 to 2015 (not shown). Syed et al. (2014) predicted that the mean surface air temperature in the monsoon season would increase from 2.5 °C to 5 °C with increasing greenhouse gas concentrations by the end of the century. Models may underestimate the rate of global warming due to the insufficient consideration of dimming and brightening (Wild, 2012). With brightening considered, warming may accelerate in South Asia at unprecedented rates. However, the recent monsoon rainfall roughly weakened with no homogeneity by combining the model and observations of satellites and sites (Bollasina et al., 2011; Turner and Annamalai, 2012; Annamalai et al., 2013). Reduced clouds may indicate a decrease in average rainfall during the monsoon season, whereas brightening generally enhances the generation of rainfall. Turner and Annamalai (2012) showed consistent negative trends of rainfall over northwest India and coastal Burma, while the positive trends in southeast India; they also found a distinct disagreement among different rainfall data in northeast India. In general, the recent brightening may play a prominent role in modulating the warming and rainfall variations in the South Asian subcontinent.

4 Conclusions

Surface solar radiation is the ultimate energy source for life on the planet (IPCC, 2013). Any change in surface solar radiation profoundly affects the global ecosystem, further determines the living conditions of humans (Haywood et al., 2011). South Asia is experiencing a leveling-off trend with regard to solar radiation, even a transition from dimming to brightening. This process is significantly complicated because of the active atmospheric action during the monsoon season. In this study, we use observations from multiple satellites/sensors on the A-Train satellite to evaluate the effect of aerosol, cloud, and water vapor variations on the recent changes in solar radiation at the surface during the monsoon season (June–September) in the South Asian subcontinent from 2006 to 2015, mainly including CloudSat, CALIPSO, MODIS and CERES.

We found the SSR and OSR increased by 16.2 W m\(^{-2}\) and decreased by 13.0 W m\(^{-2}\) with a partly fluctuation during the monsoon season, respectively. On the contrary, the increase in AOD was inconsistent with the SSR variation, and did not contribute to the recent brightening in the South Asian subcontinent. Decreases in water vapor amount and clouds significantly contributed to solar brightening and subsequent surface warming in the South Asian subcontinent. In general, the clouds were generally reduced and thinned by approximately 8.8% and 280 m, respectively, when the cloud water path (by 54.0 g m\(^{-2}\)) and particle number concentration in cloud-sky condition decreased. Given the change in clouds, the atmospheric vertical SW heat rating decreased by 0.3–0.4 K d\(^{-1}\) at the altitude of 5–15 km, whereas SW heat rating slightly increased at the low-cloud and near-surface areas (below 4 km) from 2006 to 2015. Correspondingly, SW CRE weakened by approximately 42.1 W m\(^{-2}\) at the surface. Moreover, the PW in clear-sky condition decreased by nearly 2.8 mm over the brightening period. Consequently, solar brightening increased by approximately 2.5 W m\(^{-2}\) because of the weakened absorption. The decreases in water vapor amount and clouds weakened the effect of water vapor and clouds on solar
radiation, thereby resulting in the increased absorption of direct solar radiation at the surface and subsequent surface brightening.

Notably, CloudSat is insufficient in observing thin cirrus with small IWC, thereby resulting in uncertainties for quantifying changes in clouds. In a globally averaged experiment on radiative fluxes, L'Ecuyer et al. (2008) proposed that the impact of the thin cirrus that is not detected by CloudSat on SW radiation at the surface is $-1.2 \text{ W m}^{-2}$, which is much smaller than the impact of low clouds. The data in this study were from the instantaneous observations of multiple sensors. Thus, the magnitudes of radiative impact may have been overestimated for the water vapor and cloud variations. The results are more representative of the atmospheric condition variations at nominally constant solar zenith angle than the daily average. In sum, the conclusions of this study are beneficial in the evaluation of the recent variations in aerosol, cloud, and water vapor and their resultant influences on solar radiation. Additionally, the aerosol, cloud, and water vapor variations are correlated with regional sources and meteorology. Consequently, solar dimming and brightening may be of local or regional nature, and they are unavoidably inhomogeneous and unstable in space (Soni et al., 2016). The variations in aerosols, water vapor, and clouds highly interact with changes in monsoon circulation. Thus, it would be interesting to investigate the change mechanisms of aerosols, clouds, and water vapor in South Asia. The chicken-and-egg relationship between monsoon and aerosols, water vapor, and clouds can also be explored.

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<th>Sensor</th>
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</table>
Figure 1. Terrain map during the monsoon season in the South Asia and surrounding regions. The blue line indicates rivers; the study region is within the red line.
Figure 2. Temporal variations in spatial average AOD from (a) CALIPSO and MODIS during the monsoon season, as well as (b) MODIS during pre-monsoon, monsoon, and dry seasons in the South Asian subcontinent from 2006 to 2015, respectively. The dashed lines indicate the linear fit line of the solid line with the same color; $p$ is the significance level; $k$ is the slope of the line trend of each time series.
Figure 3. Temporal variations in spatial average (a) SSR and (b) OSR from CERES during pre-monsoon, monsoon, and dry seasons in the South Asian subcontinent from 2006 to 2015, respectively. The dashed lines indicate the linear fit line of the solid line with the same color; $p$ is the significance level; $k$ is the slope of the line trend of each time series.
Figure 4. Temporal variations in spatial average vertical cloud physical parameters from CloudSat during the monsoon season in the South Asian subcontinent from 2006 to 2015: (a) cloud vertical frequency distribution, (b) LWC and (e) IWC, (c) LER and (f) IER, and (d) LNC and (g) INC.
Figure 5. Temporal variations in spatial average (a) cloud fraction and CWP, as well as (b) uppermost CTH, lowermost CBH, and CGD; spatial distributions of the temporal changes in average (c) cloud fraction, (d) CH, (e) CWP, and (f) CGD from CloudSat during the monsoon season in the South Asian subcontinent from 2006 to 2015. The dashed lines indicate the linear fit line of the solid line with the same color; $p$ is the significance level; $k$ is the slope of the line trend of each time series.
Figure 6. Temporal variations in spatial average vertical (a) SW, (b) LW, and (c) net heat rating in all-sky condition from CloudSat during the monsoon season in the South Asian subcontinent from 2006 to 2015.
Figure 7. Temporal variations in spatial average (a) cloud fraction and CRE, as well as spatial distribution of the temporal changes in average (b) SW CRE at the surface and (c) SSR from CloudSat during the monsoon season in the South Asian subcontinent from 2006 to 2015. The dashed lines indicate the linear fit line of the solid line with the same color; \( p \) is the significance level; \( k \) is the slope of the line trend of each time series.
Figure 8. Temporal variations in spatial average vertical RH in (a) all-sky and (c) clear-sky condition; spatial distribution of temporal changes in average PW in (b) all-sky and (d) clear-sky condition from ECMWF-AUX during the monsoon season in the South Asian subcontinent from 2006 to 2015.
Figure 9. Temporal variations in spatial average (a) PW (blue line), as well as SSR from BUGSr (red line) and CloudSat (orange line); spatial distribution of temporal changes in average (b) SSR from BUGSr in clear-sky condition during the monsoon season in the South Asian subcontinent from 2006 to 2015. The dashed lines indicate the linear fit line of the solid line with the same color; $p$ is the significance level; $k$ is the slope of the line trend of each time series.