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Geophysical Research Letters

Supporting Information for

Satellite-based Estimation of Cloud-top Radiative Cooling Rate for Marine Stratocumulus

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Text S1: Radiative transfer model and the configuration

We use the Santa Barbara DISORT Atmospheric Radiative Transfer (SBDART) model [Ricchiazzi *et al.*, 1998]. The SBDART code ingests sophisticated discrete ordinate radiative transfer calculations and the atmospheric transmission models. It has been widely used in atmospheric researches. We specify the vertical grids with resolutions of 50 m from the surface to 2.25 km, 200 m from 2.25 km to 8 km, and 3 km from 8 km above. The surface type is set to “sea water”. The ozone profile is set to default values for the tropical ocean. The input optical depth is uniformly distributed over the altitude range of the cloud layer. The SBDART was run twice for each case: the longwave run (5 ~ 40 μm) and the shortwave run (0.1 ~ 5 μm). The increment of the wavelength range is set to 0.1 μm .

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49 **Text S2:** Influence of H_t on CTRC

50 To illustrate the H_t influence on CTRC, we take the same example in Figure 1 and vary
51 the satellite derived H_t from 1.0 to 1.5 km by varying the lapse rate from 9.7 to 6.5 K/km. Here,
52 the T_t and SST are fixed, so their effects on CTRC are effectively controlled. Figure S4a-c shows
53 the vertical profiles of temperature, moisture, and simulated heating rate for the six experiments
54 with varied H_t . Figure S4d shows the sensitivity of the CTRC to the H_t . Only weak sensitivity
55 (several $W\ m^{-2}$) is noted. Although the H_t differs by as large as 0.5 km (quadruples the mean
56 error of satellite H_t retrieval, 0.13 km, as shown in Figure S1a), the radiative cooling values are
57 close to the “ground truth” values marked by the black circles and triangles in Fig. S4d. Such an
58 insensitivity arises from the fixation of T_t . On one hand, due to the fixed T_t , the cloud infrared
59 emission remains unchanged. On the other hand, the water vapor mixing ratio at the cloud top
60 also remains nearly unchanged because it is constrained by the T_t (Fig. S4b). Being most
61 adjacent to the cloud top, the cloud-top moisture dominates the downwelling infrared radiation
62 over the moisture loading in higher altitudes, leaving the downwelling infrared radiation vary
63 little under the fixed- T_t condition.

64 There is a slight enhancement of CTRC with increased H_t . This is due to the free-
65 tropospheric precipitable water decreasing with increased H_t (Fig. S4b). This mechanism can be
66 used to interpret a potential cancellation of CTRC when the T_t is changed. For example,
67 underestimation in T_t causes underestimation in CTRC via the weakened Planck function.
68 However, a lower T_t gives a higher H_t , enhancing the CTRC and compensating for the
69 underestimated CTRC.

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71 **Text S3:** Error propagation analysis

72 Calculation of CTRC involves complicated radiative transfer equations, so it is not feasible
73 to quantify the uncertainty mathematically. Following *Wood and Bretherton* [2004], we
74 construct “erroneous” input by assigning estimated errors to each input parameter. This gives us
75 “erroneous” CTRC, validating of which against the original CTRC gives us the root-mean-
76 square error (RMSE) and bias for each input parameter. The overall error is calculated assuming
77 random errors in all the input parameters. The result of the error propagation analysis is
78 summarized in Table S1.

79 Aircraft measurements over the southeast Pacific shows that the satellite retrieved τ and r_e
80 for closed-cell have RMSE of ~ 2 and $\sim 2 \mu\text{m}$, respectively [*Painemal and Zuidema, 2011; Witte*
81 *et al., 2018*]. At nighttime, because we assign climatological mean values to τ and r_e , we estimate
82 the uncertainties to be 6 and 5 μm for τ and r_e , respectively, based on their climatological
83 standard deviations from satellite data [*Chang et al. 2007*]. Note that the dependence of CTRC
84 on τ is nonlinear: the dependence is strong in semi-transparent clouds and it saturates after τ
85 reaches ~ 10 [*Zheng et al., 2016*]. So the potential error could be much larger than the estimation
86 here for semi-transparent stratocumulus decks. For SST and T_t , we assign an error of 1K
87 according to *Wood and Bretherton* [2004]. Our validation result (Fig. S1c) shows that the
88 RMSE of reanalysis PW_{FA} is $\sim 2\text{mm}$ ($\sim 30\%$ percentage error). Here we use the percentage error,
89 instead of RMSE, to construct the “erroneous” PW_{FA} because some cases have $\text{PW}_{\text{FA}} < 2 \text{ mm}$
90 and a minus PW_{FA} makes no physical sense. For the lapse rate of marine boundary layer, we
91 assign an error of 1 K/km, which corresponds to 0.1~ 0.2 km error in H_t for typical marine
92 boundary layers.

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101 **Table S1:** Propagated bias and RMSE in Δ_{CTF} and Δ_{BLF} for estimated errors (second column) in
 102 each input parameter. The overall errors in bias and RMSE (last row) are calculated assuming
 103 random errors in all input parameters. The first and second value in the parenthesis of the last
 104 row represents the daytime and nighttime, respectively.

Input parameter	Error	Δ_{CTF}		Δ_{BLF}		
		(W m ⁻²)		(W m ⁻²)		
		Bias	RMSE	Bias	RMSE	
τ	Daytime	± 2	0.9	4.0	-1.4	4.3
	Nighttime	± 6	5.3	11.6	0.0	4.5
r_e	Daytime	$\pm 2 \mu\text{m}$	0.3	1.9	-1.7	3.8
	Nighttime	$\pm 5 \mu\text{m}$	1.1	3.5	-2.3	3.9
SST	$\pm 1 \text{ K}$	0.1	4.5	-2.2	5.1	
T_t	$\pm 1 \text{ K}$	0.0	4.4	-2.1	5.5	
PW_{FA}	$\pm 2 \text{ mm } (\pm 30\%)$	-0.5	6.9	-2.4	7.6	
Lapse rate	$\pm 1 \text{ K/km}$	-0.5	4.5	-2.7	5.7	
Overall error (W m⁻²)		0.5	5.7	-2.0	5.5	
		(0.4/0.7)	(4.2/6.7)	(-1.7/-2.3)	(4.8/6.0)	

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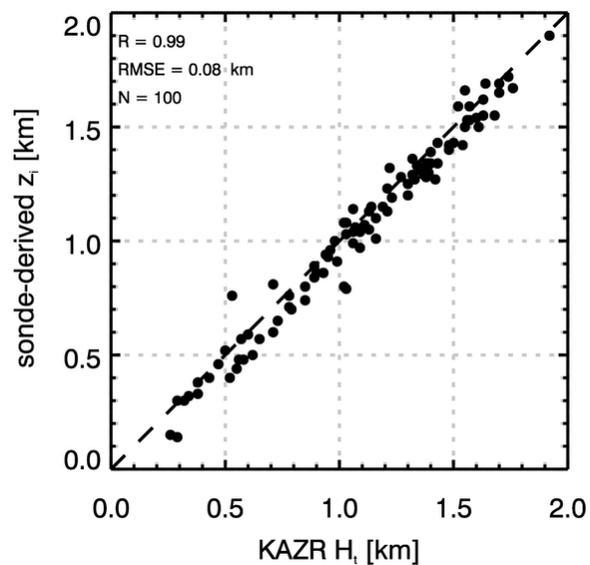
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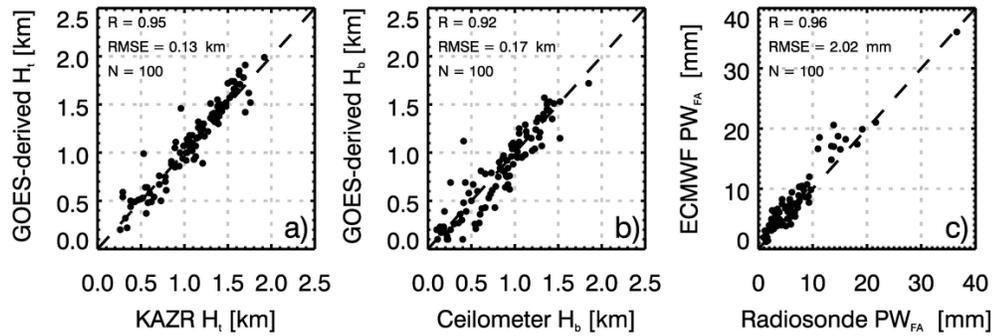
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Figure S1: Comparison between the sonde-derived z_i and the KAZR-derived H_t . The H_t represents the three-hour average of the KAZR-measured cloud top heights. We use thresholds of signal-to-noise ratio of -13 and reflectivity of -40 dBZ to identify the cloudy pixels.

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132 **Figure S2:** Comparisons between GOES-derived (a) H_t , (b) H_b , and (c) PW_{FA} against
133 ARM ground-based measurements. The H_t and H_b are three-hour averages, measured from
134 KAZR and ceilometer, respectively. Filled circles and upward triangles stand for daytime and
135 nighttime cases, respectively.

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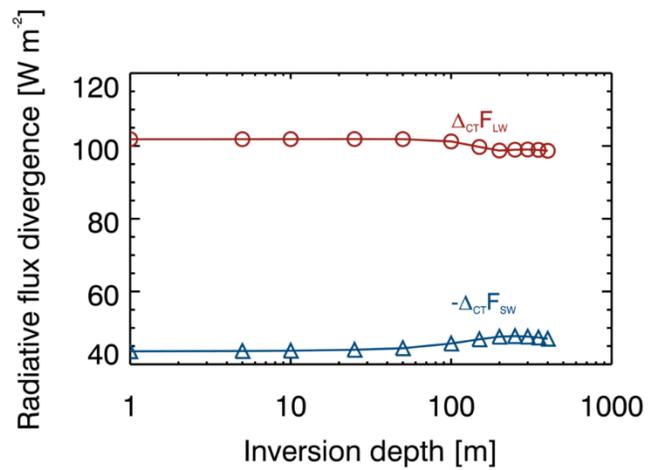
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149 **Figure S3:** Sensitivity of the $\Delta_{CT}F_{LW}$ and $-\Delta_{CT}F_{SW}$ to the depth of the inversion capping the
150 boundary layer for the composite mean case. Increase in inversion depth can increase the water
151 vapor loading in the inversion layer, thus weakening the $\Delta_{CT}F_{LW}$ slightly. The more humid
152 inversion-layer air absorbs additional shortwave radiation, increasing the $-\Delta_{CT}F_{SW}$ by just several
153 $W m^{-2}$.

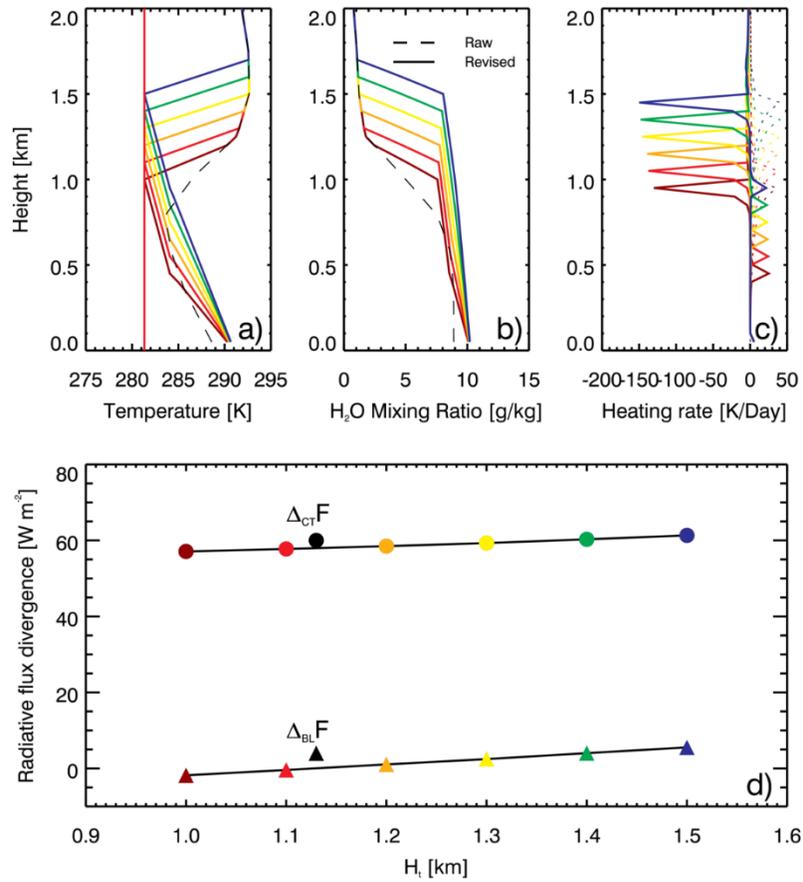
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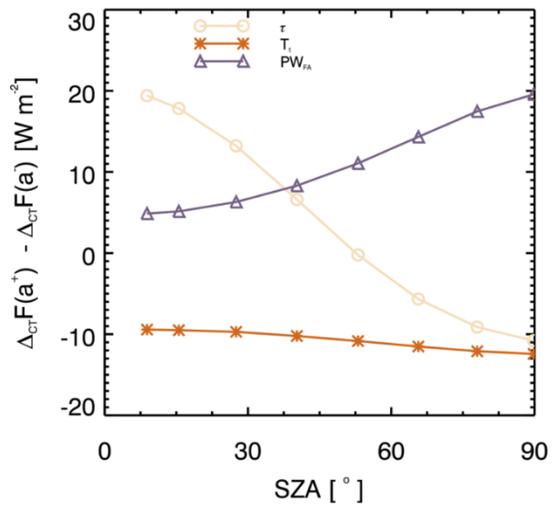
160 **Figure S4:** Vertical profiles of temperature (a), water vapor mixing ratio (b), and simulated
 161 longwave (solid) and shortwave (dotted) heating rate (c) for the six experiments with varied H_t .

162 (d) Dependence of the $\Delta_{CT}F$ (filled circle) and $\Delta_{BL}F$ (filled upward triangle) with the H_t . The

163 black symbols show the radiosonde-based estimations.

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167 **Figure S5:** Variations of the Δ_{CTF} sensitivity to the three most influential input parameters
 168 with the solar zenith angle. In the y-axis, the a in the parenthesis refers to the input parameters
 169 and the superscript “+” and “-” refer to the 75th and 25th percentiles of the composite of each
 170 input parameter.

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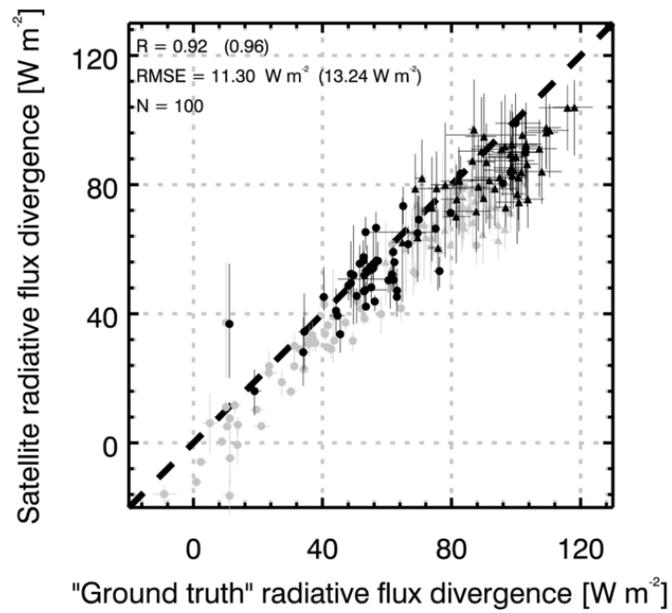
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185 **Figure S6:** Same to Figure 3, but use NCEP/NCAR reanalysis instead of ECMWF.

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