

# On the relationship between low cloud variability and lower tropospheric stability in the Southeast Pacific

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**Abstract.** In this study, we examine marine low cloud cover variability in the Southeast Pacific and its association with lower-tropospheric stability (LTS) across a spectrum of timescales. On both daily and interannual timescales, LTS and low cloud amount are very well correlated in austral summer (DJF). Meanwhile in winter (JJA), when ambient LTS increases, the LTS–low cloud relationship substantially weakens. The DJF LTS–low cloud relationship also weakens in years with unusually large ambient LTS values. These are generally strong El Niño years, in which DJF LTS values are comparable to those typically found in JJA. Thus the LTS–low cloud relationship is strongly modulated by the seasonal cycle and the ENSO phenomenon. We also investigate the origin of LTS anomalies closely associated with low cloud variability during austral summer. We find that the ocean and atmosphere are independently involved in generating anomalies in LTS and hence variability in the Southeast Pacific low cloud deck. This highlights the importance of the physical (as opposed to chemical) component of the climate system in generating internal variability in low cloud cover. It also illustrates the coupled nature of the climate system in this region, and raises the possibility of cloud feedbacks related to LTS. We conclude by addressing the implications of the LTS–low cloud relationship in the Southeast Pacific for low cloud feedbacks in anthropogenic climate change.

et al., 1984; Albrecht et al., 1988). They have long been recognized as an essential element of the global climate system owing to their cooling effect (Ramanathan et al., 1989; Hartmann et al., 1992; Klein and Hartmann, 1993). An increase in low cloud cover would attenuate global warming while a decrease would amplify it. Recognition of their essential role in climate has not yet translated into accurate simulation of marine low clouds in current climate models (Bony and Dufresne, 2005; Stephens, 2005; Clement et al., 2009). Low cloud top typically coincides with the boundary layer top, characterized by a very sharp temperature and moisture inversion. Simulating these sharp gradients explicitly may require higher vertical resolution than that typically used in current climate models. Low horizontal resolution of the models also limits the ability to simulate intense coastal atmospheric jets, which drive upwelling and cold sea surface temperature anomalies. Meanwhile, parameterization of low clouds has also proved challenging due to difficulties in representing cloud microphysical and optical processes (Bretherton et al., 2004; Teixeira et al., 2008). It may help to eventually improve low cloud parameterization and simulation by increasing our understanding of observed low clouds and controls on their variability. Various meteorological factors, including large-scale dynamics (e.g., surface divergence, circulation) and thermodynamics (e.g., SST, air-sea fluxes) have been suggested to be linked to variations in low cloud amount.

Among the various observed relationships between the ambient factors and low cloud amount, that between the seasonal variations in the strength of the temperature inversion and low cloud amount appears to be the most reliable (Stevens and Brenguier, 2009). This inversion suppresses mixing of the boundary layer and the dry free troposphere, allowing moisture evaporated from the sea surface to accumulate within the boundary layer. The moisture saturation and low cloud formation are more likely with a stronger inversion (Albrecht et al., 1988). Klein and Hartmann (1993)

## 1 Introduction

Marine low-level clouds are prevalent over eastern subtropical oceans just west of the continents, and are maintained through interactions with the lower portion of the atmosphere and the cool ocean surface (Schubert et al., 1979a; Randall



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introduced the concept of lower-tropospheric stability (LTS, defined as the potential temperature between 700 hPa and near surface,  $\theta_{700}-\theta_{1000}$ ) as a measure of the temperature inversion strength. They found that the seasonal cycle of observed low cloud cover in the main low cloud regions is closely tied to the seasonal cycle of LTS. Recent observational studies of low clouds in different low cloud regions have shown that LTS and low cloud cover are strongly correlated on seasonal timescales (Mansbach and Norris, 2007; Lin et al., 2009; Ghate et al., 2009; Kawai and Teixeira, 2010). However, It is not clear whether the seasonal cycle provides a full sampling of the atmospheric conditions shaping the relationship between the two variables. Since nearly all atmospheric variables exhibit strong seasonality, an association between the seasonal cycles of two variables such as low cloud cover and LTS may by itself be rather weak evidence for a physical mechanism linking them. Moreover, these low clouds display substantial internal variability on timescales from daily to interannual and beyond (Rozendaal et al., 1995; Klein, 1997; Garreaud et al., 2001; Rozendaal and Rossow, 2003; Xu et al., 2005; Ghate et al., 2009; George and Wood, 2010). Do we expect low cloud internal variability to exhibit the same relationship seen in the seasonal cycle? In this study, our hypotheses include that the tight association between LTS and low cloud amount on the seasonal cycle might break down on other timescales and that the LTS–low cloud relationship might be seasonally dependent and also timescale dependent. Our study is also an exploration of the degree to which and the circumstances under which the large-scale physical component of the climate system is solely responsible for internal low cloud variability. For example, it has been suggested that variability in aerosols is a significant factor generating low cloud anomalies (Albrecht, 1989; Lohmann and Lesins, 2002; Bretherton et al., 2004). If cloud anomalies are tightly associated with LTS, it is unlikely those same anomalies are generated by aerosol fluctuations.

To test our hypotheses, here we choose the Southeast Pacific as our region of interest. This region off the coast of South America from the equator to central Chile (40° S) contains one of the largest and most persistent stratocumulus decks in the world. We examine the LTS–low cloud relationship in the context of internal variability on interannual and daily timescales. A robust statistical relationship between low cloud cover and LTS anomalies would provide a strong additional line of evidence for a physical mechanism linking the two variables. On the other hand, the absence of a statistical relationship would reveal the meteorological contexts where low cloud and atmospheric stability become decoupled. As we demonstrate, such regimes are difficult to detect through analysis of the seasonal cycle alone.

The sources of LTS anomalies are ambiguous, as temperature variability both of the near-surface and above the boundary layer could lead to variations in LTS. In cases where there is a robust relationship between low cloud cover and LTS,

and LTS anomalies can be said to cause low cloud variability, this ambiguity indicates that low cloud variability could have either an atmospheric or oceanic origin. To the degree the ocean generates LTS and hence low cloud variability, this points to the fundamentally coupled nature of the climate system in the world's large stratocumulus decks. It also opens up the possibility of climate feedbacks arising from the relationship between low cloud cover and LTS, since the solar radiation anomaly associated with a low cloud anomaly could eventually lead to a temperature anomaly in the surface ocean. To shed light on these issues, we also examine the origin and structure of LTS anomalies in cases where the LTS–low cloud relationship is robust.

The paper is organized as follows. The datasets are outlined in Sect. 2, followed by a presentation of low cloud cover climatology and seasonal cycle in Sect. 3. In Sect. 4, the relationship between LTS and low cloud cover on interannual timescales is discussed. Section 5 examines the low cloud variability on the daily timescales and its relationship with LTS. Comparison of the LTS–low cloud relationship between daily and interannual timescales and modulation of this relationship by the El Niño–Southern Oscillation (ENSO) are discussed in Sect. 6. Major findings and discussion are found in Sect. 7.

## 2 Datasets

Cloud amount is taken from the International Satellite Cloud Climatology Project (ISCCP) monthly D2 (July 1983–June 2002) and 3-h D1 (December 1997–November 2001) data, given on a  $2.5^\circ \times 2.5^\circ$  grid (Rossow and Schiffer, 1991, 1999). In ISCCP, clouds are defined as low-level if their tops are at a pressure greater than 680 hPa. They are further categorized by cloud optical thickness into stratus, stratocumulus and cumulus types. In the Southeast Pacific, stratocumulus predominates over the other two low cloud types. It is well-known that ISCCP has trouble locating cloud-top pressure under inversions and other problems, such as reporting high thin clouds as mid-level clouds over some land regions (Mace et al., 2006). Our study is particularly vulnerable to the fact that because some low-level clouds are obscured by higher (middle and high) clouds, satellites misclassify them into upper-level categories (Minnis et al., 1992; Rozendaal et al., 1995). Recent buoy observational studies have verified the underestimation of ISCCP observed low cloud amount (Ghate et al., 2009). By assuming low clouds are randomly overlapped with middle and high clouds (Rozendaal et al., 1995) and the fact that most ISCCP middle level clouds over the subtropical eastern oceans are really low clouds (Garay et al., 2008), we adjust the low cloud amount as:

$$L' = (L + M)/(1 - H), \quad (1)$$

$L, M$  and  $H$  denote ISCCP's low, middle and high cloud amount respectively and  $L'$  is the adjusted low cloud amount.

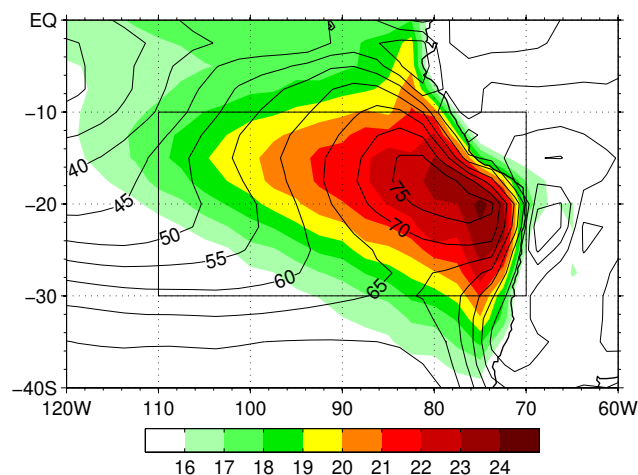
The meteorological fields used in this study are from the European Center for Medium Range Weather Forecasts Reanalysis (ERA-40) monthly (July 1983–June 2002) and 6-h (December 1997–November 2001) data, given on a  $2.5^\circ \times 2.5^\circ$  grid (Uppala et al., 2005). More modern ERA-Interim data are also used and the results are qualitatively in agreement with ERA-40 data. SST fields are from NOAA Optimum Interpolation monthly SST version 2 (OISST.v2) data, given on a  $1^\circ \times 1^\circ$  grid (Reynolds et al., 2002). Monthly data are used for analysis of seasonal and interannual variability, and sub-daily data are averaged and detrended to create a daily timeseries for analysis of day-to-day variability.

LTS is calculated from ERA-40 data as in Klein and Hartmann (1993), i.e. the potential temperature difference between the free troposphere, sampled at 700 hPa, and the near surface (1000 hPa), or  $\theta_{700}-\theta_{1000}$ . Other variants of LTS have been proposed, e.g., Estimated Inversion Strength (EIS) (Wood and Bretherton, 2006) and effective LTS (Zhang et al., 2010), to improve the relationship with low cloud when considering large-scale spatial variations spanning the subtropics and midlatitudes (Bretherton and Hartmann, 2009). We repeated analysis with EIS and results are qualitatively very similar.

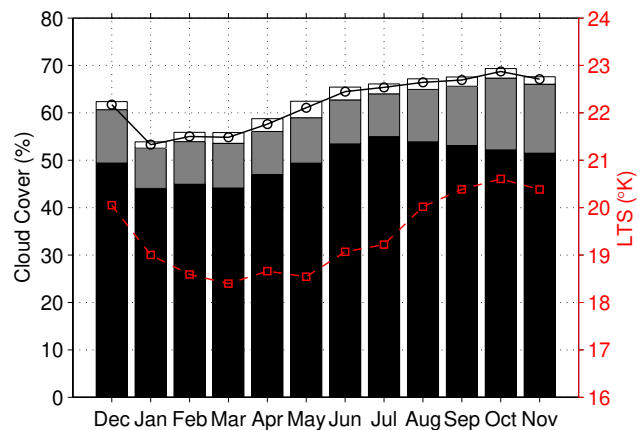
### 3 Low cloud climatology and seasonal cycle

Figure 1 displays the annual-mean adjusted low cloud amount over the Southeast Pacific, with the annual-mean LTS climatology superimposed. Low cloud amount exceeds 60 % over a large region with a spatial extent of roughly two thousand km, and approaches 75 % near Peruvian and northern Chilean coast. It decreases toward the equator and midlatitudes, as higher clouds associated with deep convection to the north and deep frontal clouds to the south predominate. This spatial pattern generally matches that of LTS. Maximum LTS values are seen just offshore of the Peruvian and northern Chilean coast. LTS then decreases westward, equatorward and poleward. The low cloud maximum is several hundred km northwest of the LTS maximum. In light of the prevailing southeast trade winds here, this is downwind of the LTS maximum. This feature suggests low cloud is not in equilibrium with local LTS, but instead is most tightly associated with LTS values 1–2 days upwind. This is consistent with the results of mixed-layer model studies (Schubert et al., 1979b).

This massive low cloud deck is present year-round, though it exhibits a modest seasonal cycle. Figure 2 shows the seasonal cycles of adjusted and unadjusted low cloud, as well as middle and high cloud, averaged over the oceanic grids within the box shown in Fig. 1. High cloud amount is negligible throughout the year. However, the middle cloud is present in all months. In austral spring (SON) middle cloud increases somewhat due to the intrusion of frontal clouds associated with midlatitude storms. This likely obscures low



**Fig. 1.** Long-term climatology (1983–2002) of lower-tropospheric stability (LTS,  $\theta_{700}-\theta_{1000}$ , shaded, Unit: K) and the adjusted low cloud amount (contour, Unit: %).



**Fig. 2.** Seasonal cycles of area-averaged ( $70^\circ$ – $110^\circ$  W,  $10^\circ$ – $30^\circ$  S, refer to box of Fig. 1) ISCCP observed low (black bar), middle (gray bar), high (white bar), adjusted low (solid line with open circle) cloud amount and LTS (dashed line with open square).

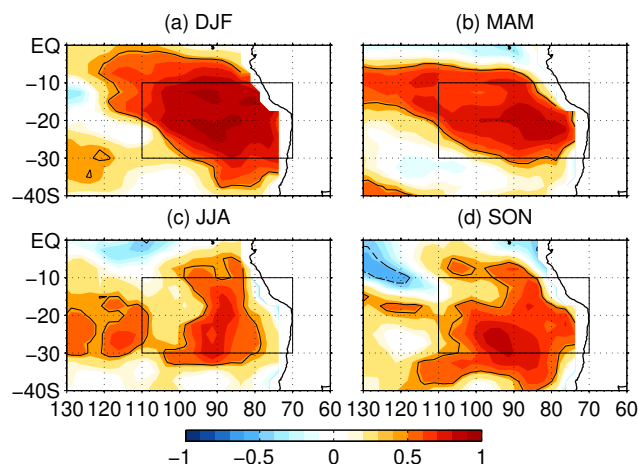
clouds, shifting the apparent maximum in unadjusted low cloud to July. The adjusted low cloud very likely gives a more realistic view of the seasonal cycle of low cloud than unadjusted low cloud. Adjusted low cloud (hereafter referred to simply as low cloud) is characterized by a persistent maximum in austral spring (SON) starting from winter (JJA), and a pronounced minimum in summer (DJF). The amplitude of this seasonal variation is roughly one eighth of the annual mean value, underscoring the persistence of the low cloud deck. The low cloud seasonality seen here is consistent in phase with that seen in Klein and Hartmann (1993), except that they reported a larger seasonality with a smaller minimum of about 40 % in DJF.

Figure 2 shows that LTS also has a distinct seasonal cycle. It rises to a peak of around 20.5 K in October and reaches its minimum of about 18.5 K in March. The seasonal cycles of low cloud cover and LTS are generally in phase, though the seasonal variation of LTS is more sinusoidal than that of low cloud, which flattens out as it reaches its maximum. This suggests the relationship between low cloud cover and LTS weakens as low cloud amount rises above about 60 %, or alternatively, LTS rises above about 19–20 K.

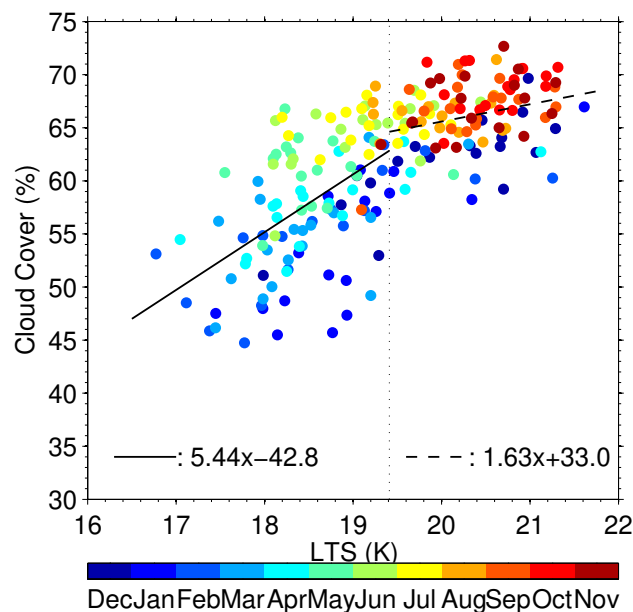
#### 4 Interannual variability

In this section, we examine the relationship between LTS and low cloud cover on interannual timescales. To simplify the analysis and focus on temporal rather than spatial variability, we will examine cloud anomalies averaged over the oceanic grids within the box in Fig. 1. To demonstrate that these area-averaged anomalies are broadly representative of coherent cloud variability in the region for each season, we correlated them with cloud anomalies at every ISCCP grid point (Fig. 3). Large and statistically significant correlations are seen at most locations within the averaging box for each season. The geographical coherence of cloud anomalies varies somewhat by season, with DJF being the most coherent across the box and JJA being the least. We further performed an Empirical Orthogonal Function (EOF) analysis (not shown) of the low cloud anomalies over the Southeast Pacific for all seasons and found that the leading modes are very similar to the patterns seen in Fig. 3 for the four seasons. This seasonal variation may reflect some seasonality in the shape of the cloud deck. In any event, Fig. 3 shows that the area-averaged anomalies are reasonable surrogates for local cloud anomalies that are in phase with one another over a large swath of the region for each season.

Having demonstrated the representativeness of the area-averaged timeseries, we show a scatterplot of monthly, area-averaged LTS versus low cloud amount, color-coded by calendar months in Fig. 4. Smaller low cloud values generally correspond to smaller LTS values, and the associations with the various months of the year are consistent with the phasing of the seasonal cycles of LTS and low cloud shown in Fig. 2. The data points in Fig. 4 behave differently depending on their relationship with the long-term LTS climatology. The points whose LTS values are less than the climatology ( $LTS < \overline{LTS}$ ) show a pronounced linear correlation between low cloud amount and LTS (correlation coefficient  $r = 0.56$ ), with a 1 K increase of LTS corresponding roughly to a 5.4 % increase in low cloud amount. Conversely, the points to the right of the dotted line ( $LTS > \overline{LTS}$ ) do not exhibit such a strong linear relationship ( $r = 0.23$ ), with the regression coefficient reduced to about 1.6 % per K significantly. Thus the association between LTS and low cloud cover practically disappears for anomalously high LTS values.



**Fig. 3.** Correlation of seasonal mean, area-averaged adjusted low cloud amount with adjusted low cloud amount at each grid point for season (a) DJF, (b) MAM, (c) JJA and (d) SON. Contour line denoting the 95 % significance level using the Student-t test is highlighted. The area of averaging is shown as a box, and is identical to the box in Fig. 1.



**Fig. 4.** Scatterplot of the monthly area-averaged ( $70^{\circ}$ – $110^{\circ}$  W,  $10^{\circ}$ – $30^{\circ}$  S, refer to the box of Fig. 1) LTS versus monthly area-averaged adjusted low cloud amount from 1983 to 2002, color-coded by calendar months. The dotted line denotes the annual mean  $\overline{LTS} = 19.4$  K. The solid (dashed) line denotes linear least-square regression line when  $LTS < (>)\overline{LTS}$ .

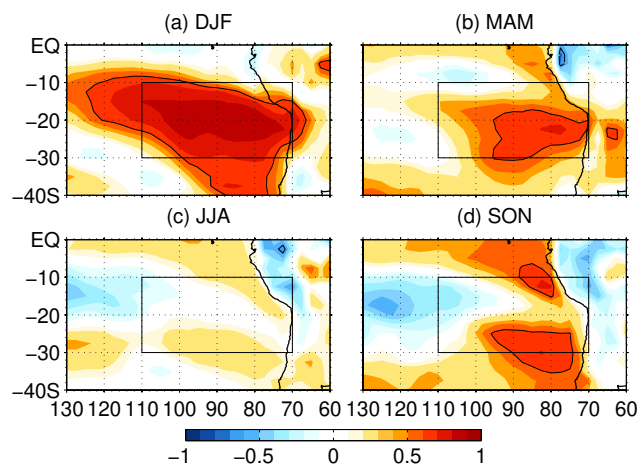
The data displayed in Fig. 4 contain signatures of both seasonally-forced and internally-generated variability. In Fig. 2, we showed that the seasonally-forced component of the variability displays a weaker relationship between LTS

**Table 1.** Slopes (Unit: % per K) of linear regression and correlation coefficients between area-averaged (70°–110° W, 10°–30° S) adjusted low cloud amount and LTS on interannual and daily timescales grouped by different seasons. Correlation coefficients with significance at the 95 % level using the Student-t test are bolded.

Slope (corr.)	DJF	MAM	JJA	SON
interannual	4.45 ( <b>0.85</b> )	3.06 ( <b>0.61</b> )	0.66 (0.19)	3.02 (0.47)
daily (2000, La Niña)	6.34 ( <b>0.78</b> )	2.48 (0.38)	1.30 (0.17)	2.41 (0.35)
daily (2001, neutral)	5.39 ( <b>0.56</b> )	0.76 (0.09)	1.33 (0.22)	2.06 (0.31)
daily (1998, El Niño)	4.97 ( <b>0.48</b> )	2.76 (0.38)	2.25 (0.38)	3.47 ( <b>0.51</b> )

and low cloud cover during austral winter and spring (JJA and SON). This could partly account for the weaker association between LTS and low cloud cover for the points to the right of the dotted line in Fig. 4. However, it turns out that the purely internally-generated component of the variability in Fig. 4 is characterized by a very similar seasonality in the LTS–low cloud relationship. We demonstrate this by calculating the seasonal means of area-averaged LTS and low cloud amount for each year and then stratifying the data by season, creating a timeseries of interannual variability for each season. Table 1 presents the resulting slopes of linear regression coefficients and correlation coefficients. In austral summer (DJF), year-to-year variations of low cloud amount are highly correlated with LTS variability ( $r = 0.85$ ). Every 1 K increase in atmospheric stability is associated with about a 4.5 % increase in low cloud amount. The fall season (MAM) exhibits a somewhat weaker relationship between stability and low cloud amount, with a smaller sensitivity of cloud amount to LTS (3.1 % per K), and a lower though still significant correlation coefficient ( $r = 0.61$ ). In winter (JJA) and spring (SON), the association between LTS and cloud becomes less significant, particularly in JJA.

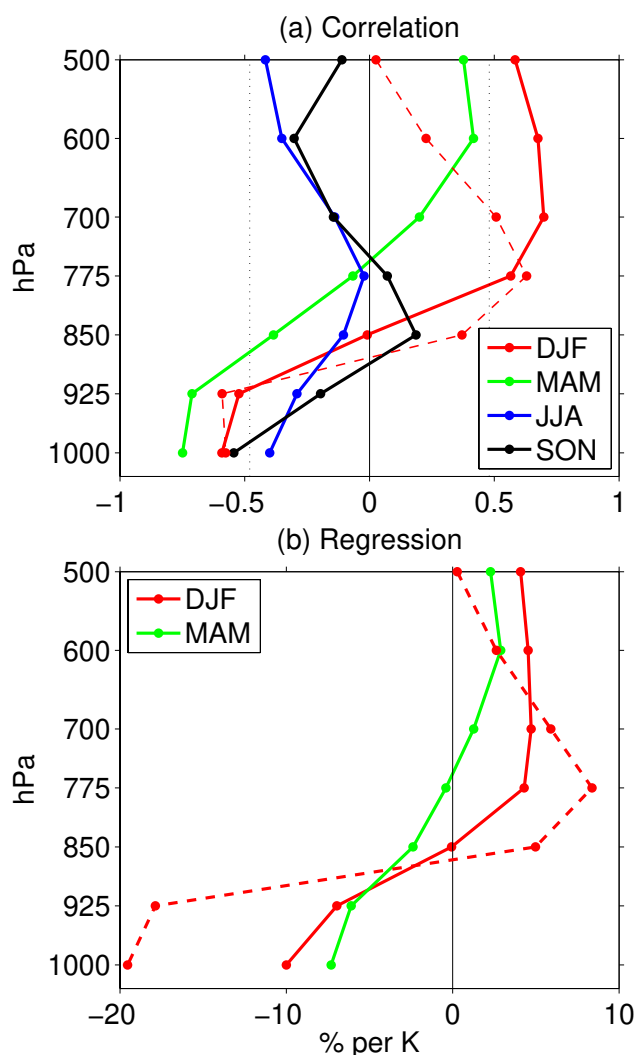
To explore the spatial relationships underpinning the LTS–low cloud statistics of internally-generated interannual variability in Table 1, we calculate for each season correlations between seasonal-mean, area-averaged marine low cloud amount and local LTS for the entire Southeast Pacific. Figure 5a displays the corresponding map for DJF. A zone of strong correlation emerges just off the coast of southern Peru and northern Chile. There is a remarkable similarity between this figure and the characteristic geographical structure of the area-averaged DJF cloud timeseries shown in Fig. 3a. This indicates that the interannual DJF variations of the entire cloud deck arise from a remarkably simple mechanism. They may be understood as a spatially-coherent response to co-located regional-scale LTS forcing. In contrast, Fig. 5b for MAM shows that significant relationships between LTS and area-averaged low cloud amount appear only in the upwind quadrant of the target area of cloud averaging, near the northern Chilean coast. If Fig. 3b is an approximate representation of the geographical structure of the area-averaged low cloud MAM timeseries, Fig. 5b suggests these cloud anomalies are generated near the Chilean coast and then advected over the



**Fig. 5.** Correlation of seasonal mean, area-averaged adjusted low cloud amount with LTS at each grid point for season (a) DJF, (b) MAM, (c) JJA and (d) SON. Contour line denoting the 95 % significance level using the Student-t test is highlighted.

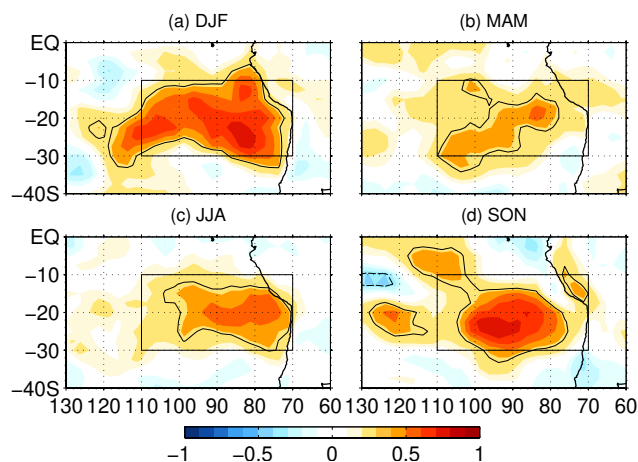
remainder of the averaging region. However, the associated LTS anomalies do not survive. Consistent with the statistics in Table 1, the correlation practically disappears for both JJA and SON for the whole Southeast Pacific (Fig. 5c and d), except for some small positive correlations in the southeast portion of the target area in SON.

Having established the horizontal structure of the LTS–low cloud relationship on interannual timescales, particularly for DJF, we next examine this relationship’s vertical structure. LTS fluctuations may arise from temperature variability either in the free atmosphere or near the surface. To determine in which part of the atmospheric column temperature variability is most closely associated with cloud variability, we show in Fig. 6a the correlation between seasonal mean, area-averaged low cloud amount and temperatures throughout the lower troposphere for the four seasons. In DJF, a positive anomaly in low cloud amount is typically accompanied by a significantly warmer free troposphere (e.g., 700 hPa) and colder near surface conditions. Both levels contribute equally to a sharper vertical temperature gradient and a correspondingly more stable boundary layer. The correlation



**Fig. 6.** Correlation coefficients (a) and regression coefficients (b) between seasonal mean, area-averaged (refer to the box of Fig. 1) adjusted low cloud amount and temperatures at each vertical level colored by seasons. Regression coefficients in only DJF and MAM are shown in (b) considering the correlation coefficients (a) are only significant through the column in those two seasons. The dotted lines in (a) denote the 95 % significance level using the Student-t test for interannual cases. The red dashed line denotes the daily analog for DJF of the year 2000.

between low cloud cover and LTS (Table 1) is higher than the correlations between low cloud cover and either constituent part of LTS, so that variability at both levels must be taken into account to maximize the association between cloud and atmospheric thermal properties. In MAM and SON, in contrast, only the near surface temperature is significantly correlated with low cloud amount, while the free atmospheric temperature is not. This suggests the MAM/SON LTS anomalies seen in Fig. 5b and d along the northern Chilean coast have



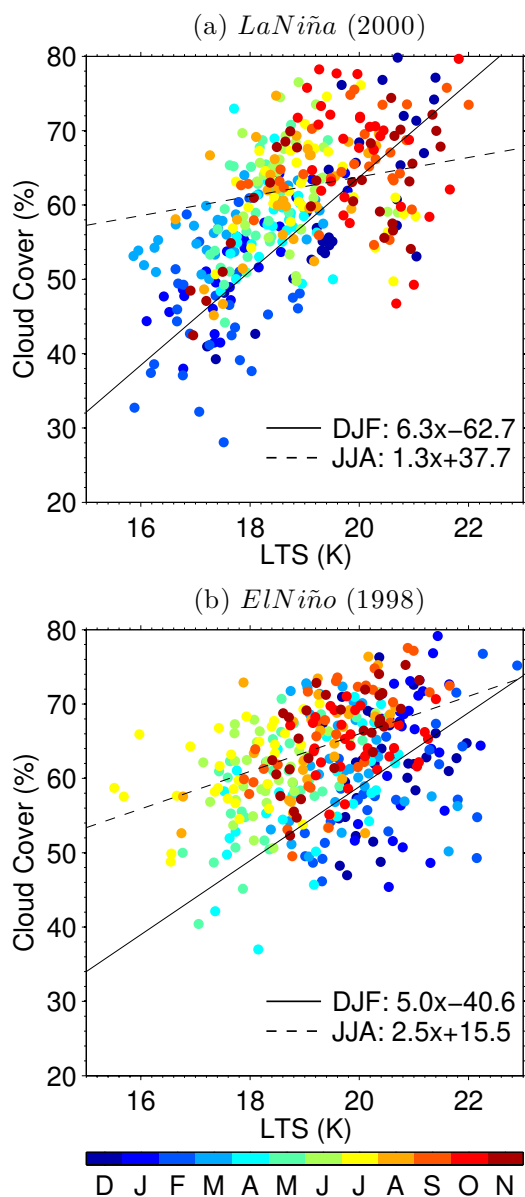
**Fig. 7.** Correlation of daily area-averaged adjusted low cloud amount with that at each grid point for (a) DJF, (b) MAM, (c) JJA and (d) SON of 2000. Contour line denoting the 95 % significance level using the Student-t test is highlighted.

their origins mainly in SST variability. A mainly oceanic origin for these MAM and SON LTS anomalies would also explain why they do not survive as the low cloud anomalies associated with them are advected into the remainder of the averaging region. Consistent with Fig. 5, low cloud amount anomalies are largely uncorrelated with atmospheric temperature anomalies throughout the lower troposphere in JJA.

The magnitudes of the cloud anomalies associated with a 1 K change in temperature at the various levels of the lower troposphere are shown in Fig. 6b. For both DJF and MAM (solid lines), the values are larger near the surface than around 700 hPa. Thus low cloud amount appears to be more sensitive to the surface temperature component of LTS. However, surface temperature variability levels are also lower than at 700 hPa. The standard deviation of DJF (MAM) temperature is 0.27 K (0.31 K) at 1000 hPa, while it is 0.68 K (0.50 K) at 700 hPa. We have seen that in DJF the sensitivity to the surface temperature component of LTS is approximately twice as large as the sensitivity to the 700 hPa component. Since the temperature variability levels are roughly twice as large at 700 hPa, the 700 hPa level may become competitive with the surface in generating a cloud anomaly simply by exhibiting more variability.

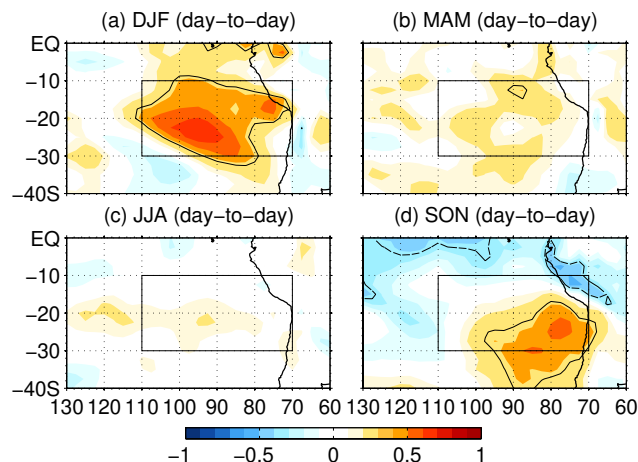
## 5 Daily variability

This section focuses on day-to-day variability in low cloud cover and its relationship with LTS. We first examine whether the area-averaged low cloud is a meaningful representation of daily low cloud variability as it is on interannual timescales. Figure 7 shows the daily analog to Fig. 3, i.e. the correlation of the daily area-averaged timeseries and low cloud amount



**Fig. 8.** Scatterplot of the daily area-averaged (refer to the box in Fig. 1) LTS versus daily area-averaged adjusted low cloud amount from all seasons of La Niña year 2000 (a) and El Niño year 1998 (b), colored by the calendar months. Solid lines denotes best-fit linear regression for austral summer (DJF) and dashed for winter (JJA) of 2000 (a) and 1998 (b).

at each grid point for each season for the year 2000. Statistically significant positive correlations are evident in each season over a large portion of the averaging area. Compared to Fig. 3, the correlations are somewhat smaller than their interannual counterparts. Low cloud day-to-day variability is less geographically coherent than interannual variability, which is associated with synoptic coastally trapped perturbations. We discuss this further in the context of Fig. 10. In general,



**Fig. 9.** Correlation of daily area-averaged adjusted low cloud amount with daily LTS at each grid point for (a) DJF, (b) MAM, (c) JJA and (d) SON of 2000. Contour line denoting the 95% significance level using the Student-t test is highlighted.

however, the area-averaged low cloud represents a large fraction of cloud variability in the region on daily timescales for each season.

We then explore whether the seasonal dependence of the LTS–low cloud relationship seen on interannual timescales also holds on daily timescales. Table 1 shows the correlation coefficients between daily mean, area-averaged low cloud cover and LTS grouped by different seasons for the year 2000. DJF stands out as the season in which day-to-day low cloud amount is best-correlated with LTS ( $r = 0.78$ ). MAM and SON shows weakened correlation ( $r = 0.38$  and  $0.35$ ) while in JJA the linear relationship disappears. This seasonal dependence is nearly identical to that seen for interannual timescales. This familiar pattern is also seen in Fig. 8a, which shows a scatterplot of daily area-averaged LTS versus low cloud cover for the year 2000, colored by calendar months and with linear least-square regression lines for DJF and JJA superimposed. This figure is qualitatively similar to its interannual variability counterpart (Fig. 4). However, for DJF daily anomalies, every 1 K change in LTS is associated with a 6.3% low cloud amount change. This is larger than the corresponding DJF value of 4.5% per K for interannual timescales. We discuss these quantitative differences in the LTS–low cloud relationship on the two timescales further in Sect. 6.

Qualitative similarities to variability on interannual timescales are also seen in the horizontal and vertical patterns of daily LTS anomalies associated with low cloud variability. Figure 9 shows the correlation map between daily area-averaged low cloud amount and LTS at each grid point for different seasons, analogous to Fig. 5. In DJF, a significant positive relationship is evident in the whole target area of cloud averaging, indicating that closely-associated

LTS and low cloud anomalies are geographically coherent over the target area. The dashed red line in Fig. 6a reveals that anomalies both near the surface and at 700–775 hPa levels contribute to the DJF LTS anomalies associated with low cloud. Daily (Fig. 6b, dashed line) DJF low cloud anomalies are more sensitive to temperatures in the free atmosphere and near the surface compared to the interannual case. As in the interannual case, the sensitivity to conditions near the surface is significantly greater than to conditions above the boundary layer. The LTS–low cloud relationship weakens in MAM and SON, with statistically significant relationships evident in only part of the target area. In JJA, the linkage between LTS and low cloud amount disappears.

Data with daily resolution allows us to examine the time evolution of synoptic atmospheric variability associated with low cloud anomalies. We do this analysis for DJF, the season with the clearest and most geographically-coherent relationship between LTS and low cloud cover. For the entire Southeast Pacific we calculate the lead-lag correlation between area-averaged low cloud amount with local temperature at the two vertical levels used to calculate LTS. Figure 10 shows the result for atmospheric temperature at 700 hPa. An anomalous warming first appears along the Peruvian and northern Chilean coasts 3–4 days ahead of a positive cloud amount anomaly. The anomalous warming then propagates westward. The propagation continues after the cloud amount peaks, until the warm anomalies move out of the target area 2–3 days later. Figure 11 shows the corresponding correlation between area-averaged low cloud cover and temperature near the surface at 1000 hPa in DJF. Anomalous near-surface cooling is persistent in a swath across the lower portion of the target area as the low clouds grow, peak, and decay. Thus the propagating signal seen in Fig. 10 is not evident here. Apparently, the timescales of variability at the two vertical levels are different. This is confirmed by a calculation of the e-folding timescales for the area-averaged temperature at 1000 and 700 hPa, 8–9 and 3–4 days respectively. The longer timescale for the near surface temperature suggests it is strongly affected by SST variability, and exhibits some of the relative sluggishness of ocean surface thermal variability (Manabe and Stouffer, 1996; Hall and Manabe, 1997; Sura et al., 2006). Furthermore, the area-averaged temperatures at 700 hPa and 1000 hPa are uncorrelated ( $r = -0.09$ ), demonstrating the complete independence of free atmosphere and near surface temperature anomalies. Thus, two independent modes of variability with different spatial and temporal structures, one primarily oceanic and the other primarily atmospheric in origin, both contribute to daily LTS variability in DJF.

## 6 ENSO modulation

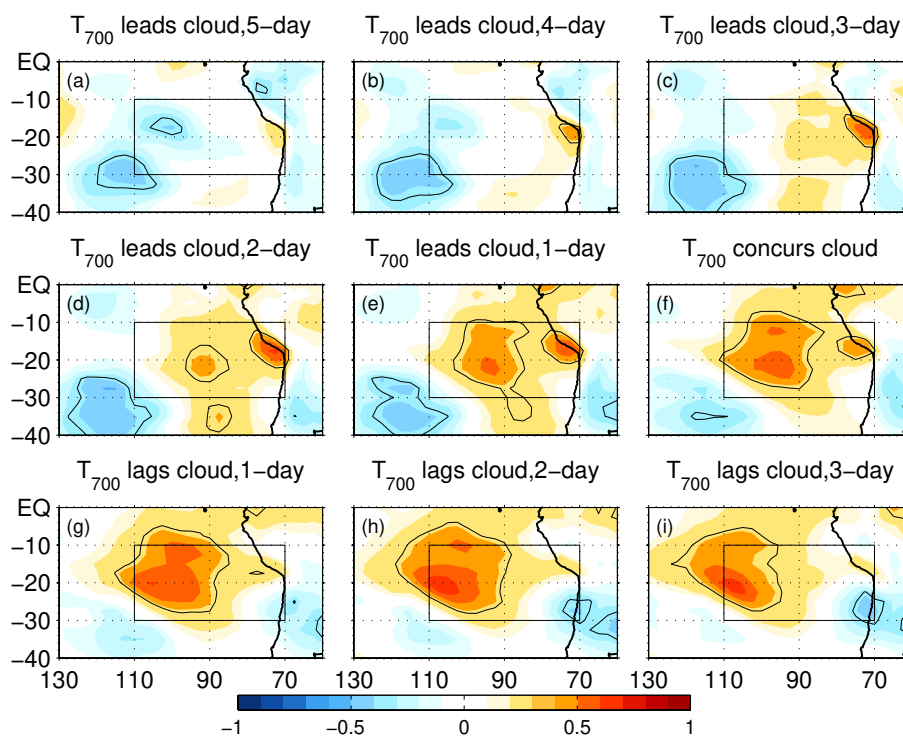
The seasonal dependence of the relationship between daily LTS and low cloud anomalies is qualitatively similar to that of the interannual anomalies discussed in Sect. 4. How-

ever, there are unexplained quantitative differences between the two timescales. For example, in DJF a given daily LTS anomaly is associated with a significantly larger low cloud anomaly than in the interannual case. It turns out that this difference can be traced to modulation of the LTS–low cloud relationship by the ENSO phenomenon. The analysis of daily variability in Sect. 5 relied on data from the year 2000. This was a La Niña year. Table 1 also shows the LTS–low cloud relationship in neutral (2001) and El Niño (1998) years. The El Niño/La Niña is defined when the 5-month running means of SST anomalies averaged in the niño3.4 region (170° W–120° W, 5° S–5° N) exceed 0.4 °C (–0.4 °C) for 6 months or more (Trenberth, 1997). The SST anomalies are calculated using the monthly OISST.v2, deviated from the 1971–2000 long-term monthly mean. Comparison of the three years indicates the relationship between LTS and low cloud amount on daily timescales depends on the large-scale state of the tropical Pacific. In DJF, a 1 K increase of daily LTS is associated with 5.0 %, 5.4 % and 6.3 % increases in low cloud amount for El Niño, neutral and La Niña years respectively. Thus low cloud anomalies associated with LTS changes are larger and low cloud cover is better correlated with LTS the further the ocean-atmosphere system is from the El Niño phase of the ENSO cycle. A visual comparison of Fig. 8a and b shows just how qualitatively different the LTS–low cloud relationship is during the two extreme phases of the ENSO cycle. During an El Niño year, the cluster of DJF points resembles the cluster of JJA points for the La Niña year, and overlaps significantly with the clusters of the other seasons.

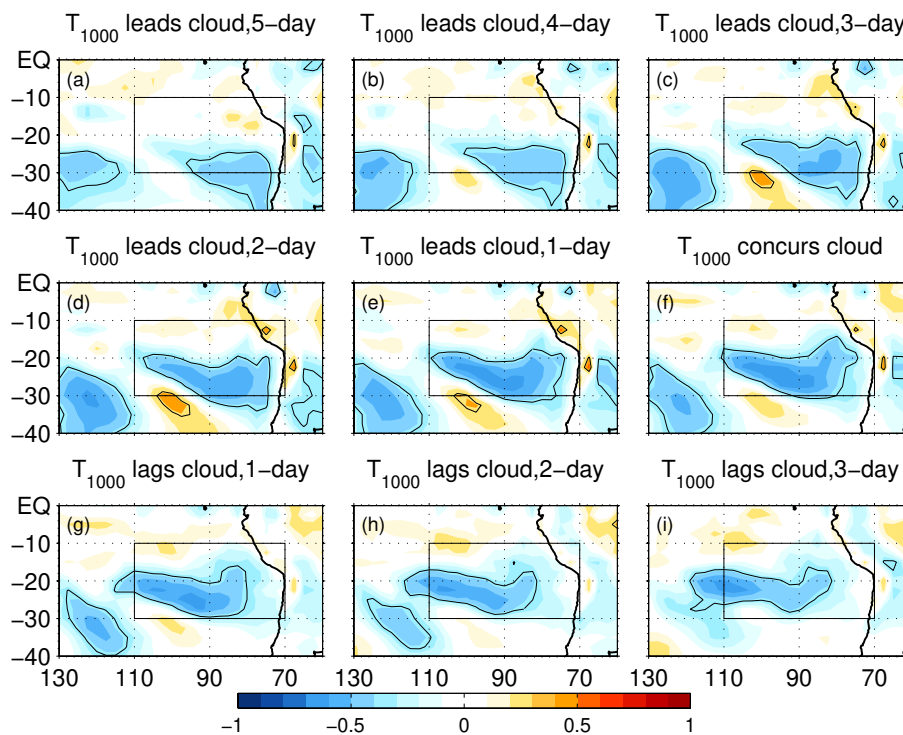
The ENSO modulation of the LTS–low cloud relationship on daily timescales raises the question of whether the imprint of the ENSO phenomenon may be seen in the LTS–low cloud relationship on interannual timescales. In Fig. 12 we show a scatterplot of the seasonal mean, area-averaged LTS versus low cloud amount for only the DJF season, colored by the DJF mean niño3.4 index. Note that the slope of the linear regression corresponding to this scatterplot is shown in Table 1. In general, the higher ENSO index years tend to be on the right side of the distributions, so that warm ENSO events are associated with higher LTS values. There are exceptions. For example, there is an El Niño event that has a relatively low LTS of around 18 K, the result of increased temperature in both the lower troposphere and the near surface. Figure 12 shows two warm ENSO events of the record, namely the 1987 and 1998 El Niño events, are visible outliers. Although these two events are characterized by very high LTS in the Southeast Pacific, the accompanying low cloud values are not correspondingly high. Including these two events significantly flattens the slope of the LTS–low cloud regression on interannual timescales. Without them, the slope of the LTS–low cloud regression increases from 4.5 % per K to 5.6 % per K, comparable to the corresponding values on daily timescales for a neutral ENSO year (Table 1, year 2001).

During a warm El Niño year, anomalous warming in the free atmosphere leads to DJF LTS values on the order of

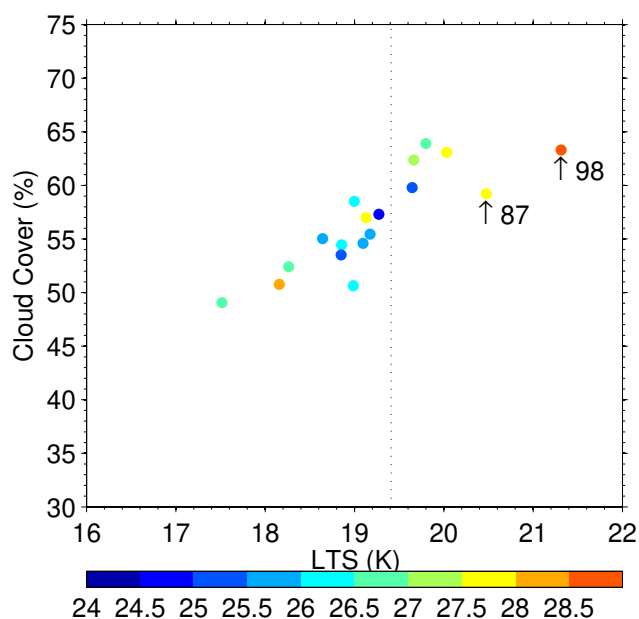




**Fig. 10.** Lead-lag correlation of daily area-averaged adjusted low cloud amount with daily temperature at 700 hPa at each grid point for austral summer (DJF) of 2000. Contour line denoting the 95 % significance level using the Student-t test is highlighted.



**Fig. 11.** Same as Fig. 10, but with temperature at 1000 hPa.



**Fig. 12.** Scatterplot of the area-averaged ( $70^{\circ}$ – $110^{\circ}$ W,  $10^{\circ}$ – $30^{\circ}$ S) LTS versus adjusted low cloud amount for DJF season from 1983 to 2002, colored by the DJF mean Niño3.4 index, an indicator of ENSO defined as area-averaged ( $170^{\circ}$ W– $120^{\circ}$ W,  $5^{\circ}$ S– $5^{\circ}$ N) SST in the tropical Pacific.

21 K, comparable to those typically found in JJA. Low cloud amount also exhibits the reduced sensitivity to LTS characteristic of JJA, reducing the slope of the LTS–low cloud regression when those points are included in the regression calculation. This is consistent with our findings in previous sections: the LTS–low cloud relationship weakens dramatically as LTS increases and reaches a threshold of about 19.5 K. This threshold, normally not breached during DJF, is very commonly breached during strong El Niño events.

## 7 Discussion and implications

In this study, we examine ISCCP observed low cloud cover variability in the Southeast Pacific and its association with ERA-40 calculated LTS across a spectrum of timescales. First, the seasonal cycles of LTS and low cloud cover are approximately in phase with each other. However, LTS shows a pronounced seasonal cycle with a distinct peak in austral spring (SON), while low cloud amount flattens out when it reaches the maximum in winter (JJA) and spring (SON). The association between LTS and low cloud cover appears to weaken in seasons with high LTS values. When considering the seasonal cycle only, every 1 K anomaly in LTS is associated with about a 5.3 % change in low cloud amount. This value is comparable to, though somewhat smaller than the values of 5.7 % and 5.0 % reported in Klein and Hart-

mann (1993) and Zhang et al. (2009). The seasonal dependence of the LTS–low cloud relationship hinted in the seasonal cycle analysis is brought into sharp relief through an examination of interannual and daily variability. On both timescales, LTS and low cloud cover are very well correlated in summer (DJF), and both LTS and low cloud anomalies are regionally-coherent. This indicates that even local low cloud variability at this time of year can be understood in terms of the regional-scale LTS–low cloud relationship. Meanwhile, though the regional-scale coherence of cloud deck variability is also present in winter (JJA), the LTS–low cloud correlation substantially weakens. The DJF LTS–low cloud relationship weakens in years with larger ambient LTS values. These years are generally strong El Niño years, in which DJF LTS values are comparable to those typically found in JJA. Thus the LTS–low cloud relationship is strongly modulated by the ENSO phenomenon.

To determine the origin of LTS anomalies closely associated with internally-generated low cloud cover variability during austral summer, we examine the relationships between low cloud variability and temperature variability at the surface and above the boundary layer. We find that both levels contribute approximately equally to low cloud variability for both interannual and daily timescales. Meanwhile, temperature variability at the two levels is essentially uncorrelated. This indicates the ocean and atmosphere are both independently involved in generating the variability of the Southeast Pacific low cloud deck, highlighting the coupled nature of the climate system in this region. The lack of correlation between the near surface and upper air was also found in observational studies of the Northeast Pacific low cloud cover (Klein et al., 1995; Klein, 1997). This result also raises the possibility of cloud feedbacks related to LTS. For example, if positive LTS and low cloud anomalies persisted long enough to substantially reduce SSTs, this would increase LTS and cloud still further. The potential importance of such a feedback is further underscored by the fact that on both daily and interannual timescales, DJF low cloud cover is significantly more sensitive to near-surface conditions than to conditions above the boundary layer. In general, our results underscore the important role of the large-scale atmospheric and oceanic conditions (i.e., the physical climate system) in generating internal low cloud variability in the Southeast Pacific.

Our results show that the LTS–low cloud relationship is only linear, and LTS is only a meaningful predictor of low cloud cover, within the regime of relatively low LTS values. The linear relationship and the predictive power of LTS practically disappear for higher LTS values. It might suggest a non-linear relationship between LTS and low cloud cover when LTS reaches certain high values. Here we suggest two potential explanations. One is that the cloud amount is not sensitive to LTS when the low cloud amount approaches larger values and becomes “saturated”. This is supported by further examination of low cloud sensitivity to LTS by dividing the box (shown in Fig. 1) and calculating averages for two

sub-regions: west of 90° W and east of 90° W. West of 90° W is a cloud transition region where stratiform cloud transitions to cumulus, while east of 90° W is the solid stratiform cloud region. Examination of low cloud's sensitivity to LTS in the two sub-regions show that the substantially weakened sensitivity to LTS is most obvious east of 90° W, where the most overcast conditions occur. West of 90° W, the transition region shows only somewhat weakened sensitivity to LTS. There have been modeling studies on this nonlinear behavior of low cloud sensitivity to LTS. For example, Zhang et al. (2009) used a mixed-layer model to study the low cloud's nonlinear response to the large-scale divergence. In their Fig. 14, they showed the enhanced large-scale divergence accompanies larger LTS, and inhibits the growth of low cloud amount. The second explanation is related to the atmospheric synoptic thermal structure important in modulating day-to-day cloud variability (Fig. 10). The synoptic activities are characterized by a westward propagation of thermal anomalies first originating from the coasts of southern Peru and northern Chile to the open ocean. These were also found in previous modeling studies (Garreaud and Rutllant, 2003) and in the recent Variability of the American Monsoon (VAMOS) Ocean-Cloud-Atmosphere-Land Study Regional Experiment (VOCALS-REx, Wood et al. (2010)) field campaign (Rahn and Garreaud, 2010). This may be attributable to zonal temperature advection induced by the land-sea thermal contrast and land-deflected synoptic storms. So the occasional offshore flow from land brings very warm and dry air over the near coastal ocean, leading to very high LTS values but conditions too dry for a cloud to form. This might also explain the substantially weakened cloud sensitivity to LTS.

Our studies also show the more regime-independent EIS has very similar relationship with low cloud cover to the LTS–low cloud relationship. The relationship between low cloud cover and LTS or EIS has been invoked in the context of current climate simulations to parameterize the low cloud amount in some weather forecast models (Slingo, 1987) and climate models (Rasch and Kristjánsson, 1998; Collins et al., 2006) and to understand low cloud cover changes in a warming climate (Stevens and Brenguier, 2009; Bretherton and Hartmann, 2009). And it has been relied on to make predictions about the behavior of low cloud feedback and the implications for future climate change. For example, Miller (1997) suggested that since LTS tends to increase in a warmer climate due to greater atmospheric warming in subtropical regions above the boundary layer than at the surface, low cloud cover will increase, leading to a negative feedback. Our results from the Southeast Pacific indicate that while it may be valid to extrapolate the low cloud cover change in a future climate where LTS increases relying on the current LTS–low cloud relationship, some caution is warranted. The exact nature of the relationship between LTS and low cloud cover appears to be regime-dependent even during the summer season when the LTS–low cloud relationship is strongest. The overall change in climate associated with an-

thropogenic forcing may be large enough to cause a change during a particular season from one regime to another. Our results also show that during austral winter and spring, most temporal variability in the low cloud deck is consistently unrelated to LTS. Other regional-scale factors must be important in generating low cloud variability, such as temperature advection, the strength of the subtropical high, the surface wind and the relative humidity of the cloud layer (Klein et al., 1995; Klein, 1997; George and Wood, 2010; Painemal and Zuidema, 2010). Besides the physical meteorological controls on low cloud, the atmospheric aerosols are also important for cloud formation, particularly the cloud microstructure (Albrecht, 1989; Bretherton et al., 2004), and potentially overall cloud amount. During these seasons, it is difficult to justify relying on the LTS–low cloud relationship to make predictions or statements about the low cloud cover response to anthropogenic forcing in the Southeast Pacific. Meanwhile, this nonlinear LTS–low cloud relationship found in the Southeast Pacific in this study may behave differently in other low cloud regions, another topic for a future study.

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