Column-integrated Moist Static Energy Budget Analysis on Various Time Scales during TOGA COARE

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ABSTRACT

Moist static energy (MSE) budgets on different time scales are analyzed in the TOGA COARE data, using Lanczos filters to separate variability with different frequencies. Four different time scales (∼2 day, ∼5 day, ∼10 day, and MJO time scales) are chosen based on the power spectrum of the precipitation and previous TOGA COARE studies. The lag regression slope technique is utilized to depict characteristic patterns of the variability associated with the MSE budgets on the different time scales.

Our analysis illustrates that the MSE budgets behave in significantly different ways on the different time scales. On shorter time scales, the vertical advection acts as a primary driver of the recharge-discharge mechanism of column MSE. As the time scale gets longer, in contrast, the relative contributions of the other budget terms become greater, and consequently, on the MJO time scale all the budget terms have nearly the same amplitude. Especially, our results indicate the horizontal advection plays an important role in the eastward propagation of the MJO during TOGA COARE. On the MJO time scale, the export of MSE by the vertical advection is in phase with the precipitation. On shorter time scales, the vertical velocity profile transitions from bottom-heavy to top-heavy, while on longer time scales, the shape becomes more constant and similar to a first baroclinic mode structure. This leads to a more constant gross moist stability on longer time scales, which we estimate.
1. Background

In order to investigate the relationship between tropical convection and its associated large-scale circulations, past work has examined column-integrated moist static energy (MSE) budgets. These budgets tell us about the processes associated with the growth and decay of column MSE. The column MSE is useful as a diagnostic quantity in the deep tropics primarily for two reasons. First, it is approximately conserved in moist adiabatic processes, and it is often beneficial to study any phenomenon from a perspective of conserved variables. Second, the column MSE is tightly connected to tropical convective variability. Column water vapor is known to be closely linked to precipitation anomalies in the tropics (e.g., Raymond 2000; Bretherton et al. 2004; Neelin et al. 2009; Masunaga 2012), and temperature anomalies are small due to the large Rossby radius (Charney 1963, 1969; Bretherton and Smolarkiewicz 1989; Sobel and Bretherton 2000). Together, these two constraints mean that the evolution of column MSE is closely related to the evolution of precipitation anomalies. In this work, we explore the charging and discharging mechanisms of column MSE that are associated with precipitation anomalies for various frequencies of variability. To do this, we examine column MSE budgets using data from the Tropical Ocean Global Atmosphere Coupled Ocean-Atmosphere Response Experiment (TOGA COARE; Webster and Lukas 1992) field campaign.

The column-integrated MSE budget equation is, following Yanai et al. (1973),

\[
\frac{\partial \langle h \rangle}{\partial t} = -\langle \vec{v} \cdot \nabla h \rangle - \langle \omega \frac{\partial h}{\partial p} \rangle + \langle Q_R \rangle + SF
\]

where \( h \equiv s + Lq \) represents MSE, \( s \) represents dry static energy (DSE), \( L \) represents the latent heat of vaporization, \( q \) represents specific humidity, \( Q_R \) represents radiative heating rate, \( SF \) represents surface fluxes of MSE, the other terms have conventional meteorology meanings and we have neglected a residual due to ice processes. The angled brackets represent a vertical integral over mass in the troposphere. Because in the deep tropics variations in the temperature field are much smaller than those of moisture, variations in \( h \) are pri-
marily due to fluctuations of atmospheric moisture. Thus investigating the column $h$ budget leads us to understand how moisture anomalies amplify and decay in the tropics.

Episodes of organized deep convection in the tropics are thought to generally begin with bottom-heavy diabatic heating\(^1\) that progressively deepens as the convection develops and eventually becomes top-heavy and stratiform. This structure has been seen in convectively-coupled equatorial waves (e.g., Takayabu et al. 1996; Straub and Kiladis 2003; Haertel and Kiladis 2004; Haertel et al. 2008; Kiladis et al. 2009), the MJO (e.g., Lin et al. 2004; Kiladis et al. 2005; Benedict and Randall 2007; Haertel et al. 2008), and even individual mesoscale convective systems (e.g., Mapes et al. 2006). The vertical profile of convection also has a strong impact on numerical simulations of the MJO (e.g., Lin et al. 2004; Fu and Wang 2009; Kuang 2010; Lappen and Schumacher 2012, 2014), convectively-coupled waves (e.g., Cho and Pendlebury 1997; Mapes 2000; Kuang 2008) and convective organization in general. These phenomena are presently very challenging to simulate correctly, which makes numerical weather prediction difficult (e.g., Lin et al. 2006; Kim et al. 2009; Benedict et al. 2013).

Interestingly, bottom-heavy profiles of vertical motion are associated with the import of MSE by the vertical circulation (i.e., negative $\langle \omega \partial h / \partial p \rangle$). These tend to coincide with the build-up of moisture in disturbances. Conversely, top-heavy profiles of vertical motion are associated with the export of MSE by the vertical circulation and these tend to coincide with the decay of moisture in disturbances. This suggests that, as pointed out by Peters and Bretherton (2006), the vertical advection term could be playing a role in the charging and discharging of column MSE associated with disturbances. This was also seen to some degree in recent work on the MSE budget during the Dynamics of the Madden Julian Oscillation (DYNAMO) field campaign (Sobel et al. 2014). In this work, we systematically examine the relative contribution of this vertical advective term, as well as other terms to the build-

\(^1\)Since most of the diabatic heating is balanced by vertical DSE advection and profiles of the DSE are relatively constant in the tropics, structures of the diabatic heating are similar to those of the vertical velocity profiles.
up and decay of column MSE for various frequencies of variability observed during TOGA COARE.

We also examine hypotheses about MJO dynamics that have been emerging from the most recent MJO studies (e.g., Kim et al. 2014; Sobel et al. 2014). That is, 1) the radiative heating and surface fluxes destabilize the MJO disturbance by amplifying and maintaining MJO MSE anomalies while 2) the vertical advection stabilizes the disturbance by exporting MSE, and 3) the horizontal advection plays a significant role in the eastward propagation by building up moist conditions ahead, and providing dry conditions behind the active convective phase. These points are investigated in the MJO events during TOGA COARE.

Neelin and Held (1987) introduced a normalized version of the vertical advective term, known as the gross moist stability, which "provides a convenient way of summarizing our ignorance of the details of the convection and large-scale transients." Other versions of this quantity have been used in many studies (see a review paper by Raymond et al. 2009). In this work, we examine the implications of the bottom-heavy to top-heavy evolution of vertical motion profiles for the gross moist stability. We also briefly discuss an appropriate choice of time filters for investigating relatively high frequency variability in the TOGA COARE data set.

Section 2 describes our data and filtering, regression methodology. In section 3, we show column-integrated MSE budgets for various time scales of variability, as well as vertical motion profiles. Section 4 has a discussion of gross moist stability and calculations of this quantity. In section 5, we discuss the relationship between a constant gross moist stability and the vertical motion structure being well-described by a first baroclinic mode. In this section, we estimate the gross moist stability in a different way from section 4 and also briefly discuss sensitivity to our filter choice. In section 6, we describe our conclusions.
2. Data and Methodology

a. Data description

We investigated the data associated with the column-integrated moist static energy budget equation during the Tropical Ocean Global Atmosphere Coupled Ocean-Atmosphere Response Experiment (TOGA COARE; Webster and Lukas 1992). TOGA COARE is a package of various field experiments conducted in the western equatorial Pacific. The experiment provided detailed observations of the mean and transient states of the tropical variability in the western Pacific warm pool, enabling identification of the dominant dynamical and thermodynamic processes in large-scale tropical convective systems. We utilized the data during the Intensive Operative Period (IOP) starting from 1 November 1992 to 28 February 1993 with 6 hourly time resolution. Each variable was averaged over the spatial domain called the Intensive Flux Array (IFA; see Fig. 14 in Webster and Lukas 1992).

The data set we used was objectively constructed by Minghua Zhang, who used constrained variational analysis for producing each variable. That method guarantees the conservation of the column-integrated mass, water, and DSE. See Zhang and Lin (1997) for more detailed description about the constrained variational analysis.

b. Selection of time scales

For examining the column MSE budgets and associated terms for different frequencies of variability, we chose four time scales, \( \sim 2 \) day, \( \sim 5 \) day, \( \sim 10 \) day, and MJO (\( > 20 \) day) time scales. Those time scales are chosen based on a power spectrum of the precipitation during TOGA COARE and previous TOGA COARE studies. Figure 1a shows the power spectrum of the precipitation. Since the purpose of this study is not to investigate spectral signals which have been already examined by many previous studies, we will not look at statistical robustness of the signals in the power spectrum. We will use this power spectrum just for the purpose to determine which time scales should be separated to be investigated.
Figure 1a shows there are four peaks with different periodicities. The first one is the diurnal cycle which is not of our interest in this study, thus was removed by filtering in the analysis. The second peak can be found around 2 day period. This signal has been investigated by Takayabu et al. (1996) and Haertel and Kiladis (2004) who have pointed out that there exist westward propagating 2-day inertia gravity waves during TOGA COARE. Thus we dealt with this time scale separately. The other signals are found around 4~5 day and 10~13 day periods, which could be Kelvin wave signals. Because those two are obviously distinct and different from the 2-day wave signal, we also examined those time scales separately. Because the signal of 10~13 day period in the power spectra is much smaller than the other signals, we cannot negate the possibility that the signal here is just a statistical noise. Nevertheless, we investigate this signal in order to keep consistency with Mapes et al. (2006), who have also investigated this periodicity in the TOGA COARE data set. Finally, the MJO time scale was extracted because many previous studies have shown there are two MJO events during TOGA COARE (e.g., Velden and Young 1994; Lin and Johnson 1996; Yanai et al. 2000; Kikuchi and Takayabu 2004) in late November to December (around 30 to 65 COARE day) and in February (around 70 to 100 COARE day). Because the second MJO signal was attenuated before reaching the IFA (see Fig.3 in Yanai et al. 2000), most of the features in the following analyses on the MJO time scale reflect the structures of the first MJO event.

c. Filtering

In order to extract different time scale features, a Lanczos filter was utilized. This filter has been popularly used in meteorology and other areas because the responses of frequencies to the filter has been well-studied (Duchon 1979) and it has desirable behaviors with minimum Gibbs oscillations and relatively sharp cut-off slopes which prevent frequencies of interest from being contaminated by undesirable leakage of frequencies and artificial false responses produced by the Gibbs oscillations. We will briefly discuss sensitivities of the
results to the choice of filtering in section 5d, where we will compare the Lanczos filter with a running mean filter, especially on short time scales.

There is a common trade-off between the number of weightings, or the number of data points which have to be sacrificed, and desirable behaviors of the filter. We chose 151 as the number of the weightings for all the analyses. This number was chosen in such a way that the response function of the filter looks appropriate enough to separate the MJO signals from the other shorter time scale signals (see Fig. 1b). Although we could have used a smaller number for the analyses on the shorter time scales (\(\sim 2\) day and \(\sim 5\) day scales) for reducing sacrificed points, we used the same number for all the analyses. We tried different numbers of weightings, and found those didn’t make significant changes in the results. Figure 1c shows time-series of the raw and filtered precipitation. We can see one strong MJO signal from around 30 November 1992 to 3 January 1993 (from 30 to 65 COARE day) and one weak signal from around 8 January to 7 February 1993 (from 70 to 100 COARE day).

\(d\). Regression analysis and correlation test

Variability on the different time scales was plotted using a linear lag-regression analysis. This method has been used by many studies (e.g., Kiladis and Weickmann 1992; Mapes et al. 2006). In this analysis, a predictand is regressed against a predictor (or a master index) to determine regression slopes at different lag times. These computed regression slopes are scaled with one standard deviation of the predictor so that the computed regression slopes have the same unit as that of the predictand. We chose precipitation as the predictor, and each variable in Eq. 1 as a predictand. We also computed the vertical structures of the regression slopes of vertical pressure-velocity (omega), wind divergence, and specific humidity on the different time scales as in Mapes et al. (2006). Those slopes were computed at each lag time and each height. Both the predictor and predictands were filtered with a Lanczos filter for statistical correlation tests. (For a regression analysis, predictands don’t need to be filtered.)
Statistical correlation tests were applied to test whether a given feature is statistically significant. Degrees of freedom (DOF) for the correlation tests were estimated at each lag and height following Bretherton et al. (1999). Although the values of the estimated DOF vary among different grids and variables, those variations are small enough that we neglect them. The DOF on \(\sim 2\) day time scale is about 102 (this is an average value of the different values of the DOF) and the DOF on \(\sim 5\) day time scale is about 22. On \(\sim 10\) day time scale, the number of different realizations (convection) can be counted in Fig. 1c and it is about 6, thus the DOF for the correlation test on this time scale is 4. For the MJO time scale, there are only two independent events. Since those numbers of the independent samples on 10 day and the MJO time scales are too small to do statistical tests, statistical significance was tested only on \(\sim 2\) day and \(\sim 5\) day time scales.

3. Results: Column MSE budgets and omega profiles

a. Column MSE budgets

In the top panels of Fig. 2, plotted are lag auto-correlations of precipitation, lag correlations between precipitation and column-integrated MSE, and in the bottom panels lag regression slopes of each term in Eq. 1 regressed against the precipitation and scaled with one standard deviation of the precipitation on the different time scales. The standard deviations of raw data, \(\sim 2\) day, \(\sim 5\) day, \(\sim 10\) day, and MJO time scales, are respectively 229 W m\(^{-2}\), 112 W m\(^{-2}\), 91 W m\(^{-2}\), 121 W m\(^{-2}\), and 123 W m\(^{-2}\). Every variable is filtered with a Lanczos filter on the corresponding time scales. Confidence intervals of the 90% significant level of the regression slopes are also plotted on the left bottom corners on only \(\sim 2\) day and \(\sim 5\) day time scales; the time scales on which we can get enough DOF. The values of confidence intervals differ at different lags, thus average values among the lag time windows are plotted. The numbers on the right corners of each subplot are average values (among the lag time windows) of the numbers of the independent samples. Increased errors on \(\sim 5\)
day time scale compared to ~2 day time scale are primarily due to the reduced DOF.

We first acknowledge that due to the lack of DOF we are uncertain about whether or not Figs. 2c and 2d represent statistically significant features of the MSE budgets on those time scales. To examine statistical significance on those time scales, we need to investigate longer time-series than the TOGA COARE data, which is left for future work. Nevertheless, we can see that the patterns in Fig. 2d for the MJO events during TOGA COARE are similar to those in Fig. 10 in Benedict et al. (2014) in which 10 year long ERA-interim and TRMM with objectively analyzed surface flux data were investigated.

Column-integrated radiative heating \( \langle Q_R \rangle \) is approximately in phase with the precipitation (or the precipitation leads slightly) on all the time scales. Surface fluxes \( SF \) lag the precipitation peaks on all the time scales except for \(~10\) day scale on which both radiative heating and surface fluxes are nearly in phase with the precipitation. The lags of \( SF \) are significant on \(~5\) day and MJO time scales (\(>20\) day).

The behaviors of column-integrated vertical MSE advection (or \(-\langle \omega \partial h/\partial p \rangle \)) differ among the time scales. On \(~2\) day scale, positive advection (i.e., import of \( h \)) leads the precipitation and the minimum value (i.e., maximum export of \( h \)) lags the precipitation peak. The tendency of column-integrated \( h \) (or \( \partial \langle h \rangle / \partial t \)) agrees with the vertical advection term, which implies that on this time scale most of the recharge-discharge cycle of \( h \) is explained by the vertical advection while the other terms cancel out each other.

On the \(~5\) day scale, the pattern of vertical advection term is similar to that of \(~2\) day scale in which positive advection leads the precipitation and negative advection lags the precipitation peak. Unlike the \(~2\) day scale, there is a lag between the vertical advection and tendency term on this time scale which is due to negative contributions of the radiative heating and surface fluxes in the early stage of the convection. This lag between the vertical advection and tendency term becomes larger as the time scale gets longer.

On the \(~10\) day scale, the maximum vertical advection leads the tendency maximum by around 3 days. Furthermore, the relative amplitude of vertical advection to the tendency
term becomes greater on this time scale, which is due to the other terms that work in the
opposite way to the vertical advection. That is, in the early stage of the convection the
vertical advection recharges the $h$ while the other terms discharge the $h$ and in the mature
stage the vertical advection exports the $h$ while the other terms recharge it.

On the MJO time scale, akin to $\sim 10$ day scale, the positive vertical advection leads the
positive tendency term and amplitude of the vertical advection is greater than that of the
tendency term because the other terms play significant roles in the $h$ budgets. It is also
worthwhile to note that as the time scale gets longer the vertical advective export of MSE
(i.e., positive $\langle \omega \partial h/\partial p \rangle$) becomes more in phase with precipitation peak (i.e., the lag relation
becomes closer to 180 degree out of phase). On $\sim 2$ day and $\sim 5$ day time-scales the vertical
advective $h$ export lags the precipitation peak, while on $\sim 10$ day and the MJO time scale
it becomes more in phase with the precipitation peak. This in-phase $h$ export pattern has
implications when we consider the gross moist stability (GMS), which will be discussed in
section 4b.

The horizontal advection (i.e., $-\langle \vec{v} \cdot \nabla h \rangle$) exhibits significantly different behaviors among
these different frequencies. On $\sim 2$ day scale, the positive horizontal advection leads the
precipitation and the minimum value reaches slightly after the precipitation peak. The hor-
izontal advection acts in almost opposite ways to the radiative heating and surface fluxes.
As a result, those terms cancel out each other. On $\sim 5$ day scale, the horizontal advection
is almost 90 degree out of phase with the precipitation. In contrast, on the $\sim 10$ day scale it
is almost in phase with the precipitation. Again, since Fig. 2c contains only 6 independent
samples we cannot conclude that this pattern is statistically robust. More detailed investiga-
tions should be done on this time scale in future work. On the MJO time scale, the horizontal
advection is 90 degree out of phase with the precipitation. Before the precipitation peak the
horizontal advection imports $h$ while after the precipitation maximum it exports the $h$. As
the time scale gets longer, the amplitude of the variations of the horizontal advection be-
come greater, which might indicate that the relative contribution of the horizontal advection
to the recharge and discharge of the MSE becomes more important as the time scale gets longer.

The relative amplitudes of each term indicate which terms are the most important for these frequencies. For all the frequencies except for the MJO, the vertical advection dominates the other terms which implies that the vertical advection is the most important sink and source. At longer time scales of variability (lower frequencies), however, the amplitude of the vertical advection term relative to the source/sink terms becomes less. On the MJO scale, the horizontal and vertical advection, radiative heating, and surface fluxes all have relatively similar amplitudes. That indicates that all the terms in the MSE budgets play important roles in the MJO dynamics.

Furthermore, the results shown in Fig. 2d on the MJO time scale reinforce the view of the MJO dynamics which has been emerging from recent studies (e.g., Kim et al. 2014; Sobel et al. 2014). That is, 1) the radiative heating and surface fluxes amplify and maintain the MJO MSE anomalies while 2) the MJO disturbance is stabilized by the vertical advection which exports MSE and cancels the effect of the radiative heating and surface fluxes, and therefore 3) the eastward propagation of the MJO is primarily driven by the horizontal advection which provides moistening ahead (in the negative lags, or to the east of), drying behind (in the positive lags, or to the west of) the active convective phase. Although there are differences between the different MJO events as pointed out by Sobel et al. (2014), our results, in general, show significant consistencies with the results given by Kim et al. (2014) and, to some degree, with the results in Sobel et al. (2014).

b. Omega profiles

Figure 3 shows vertical structures of vertical pressure velocity (omega) and wind divergence on the different time scales. The areas surrounded by the green curves passed statistical correlation tests with 99% (on ~2 day time scale) and 80% (on ~5 day time scale) significant levels. The lower significant level used on ~5 day time scale is because of smaller DOF on
this time scale compared to \sim 2 day time scale. The statistical tests were not applied for
\sim 10 day and the MJO time scales due to the lack of DOF. As Mapes et al. (2006) showed,
we can observe tilting structures of the omega profiles in which the profile evolves from a
bottom-heavy shape into a top-heavy shape (indicated by the black dash lines), and these
tilting structures are statistically significant. The figures of the wind divergence illustrate the
same information as the omega figures. Height of the lower tropospheric convergence (blue
shaded contours) rises as the convection develops, making the tilting divergence profiles.

However, one can notice that the tilt of the omega profile becomes steeper as the time
scale gets longer. Especially, on the MJO time scale, the contour line of the omega is
almost perpendicular to the isobaric surface at \sim 10 lag day. There is a shallow convective
phase on this scale, too (see from \sim 22 to \sim 12 lag days), but this shallow convection is
more abruptly changed into deep convection compared to those on the shorter time scales in
which the transitions of the convection from a bottom-heavy to a top-heavy shape happen
more gradually. The divergence figures depict the differences among the time scales clearly.
In the upper troposphere, the structures are qualitatively similar among the different time
scales. In the inactive stage of the convection, strong convergence associated with upper
tropospheric descending motion happens at the top of the troposphere. In the mature stage
of the convection, in contrast, strong divergence due to deep convection happens.

In the lower half of the troposphere, differences among the time scales are prominent.
On all the time scales except for the MJO time scale, in the inactive convective stage,
the strongest divergence happens around 600 hPa. On the MJO time scale, in contrast,
the divergence at this level is much weaker than that on the shorter time scales, and the
strongest divergence happens around 900 hPa. This lower tropospheric divergence maintains
its strength until \sim 15 lag day. As this lower tropospheric divergence disappears, the convec-
tion abruptly changes into deep convection. Therefore, on the MJO time scale, the omega
profiles behave like a single deep convection mode which is often called a first baroclinic
mode. This omega behavior has implications regarding the gross moist stability (GMS) of
Before going to the next section, it should be emphasized again that the results shown in Figs. 3g and h reflect only two MJO events, one of which is a weak event, and thus it is almost a case study. Therefore, it is difficult to draw a general conclusion about the MJO structures from our analysis particularly because the details of the MJO structures differ significantly from event to event. However, we can at least claim that a strong tilt of the omega profile (or latent heating profile) is not necessary for the existence of the MJO even though the tilt might play a role in the MJO dynamics.

Furthermore, it should also be noted that our lag-regression methodology extracted the actual structures of the MJO event during TOGA COARE in an appropriate way. Figure 4 shows the time-height plot of the anomalous omega of the first MJO event during TOGA COARE, which occurs between \( \sim 30 \) COARE day and \( \sim 65 \) COARE day. In this plot, we simply utilized a 15-day running mean filter. Although the contour is noisy due to the noise introduced by the running mean filter, the overall structure is similar to that in Fig. 3g. This figure indicates that our methodology captures the MJO structures well, and negates the possibility that the result shown in Fig. 3g is due to a false signal introduced by the statistical method.

4. More results: Gross moist stability

a. GMS with different frequencies

Now the gross moist stability (GMS) on the different time scales will be computed. Before doing actual computations, the concept of the GMS needs to be clarified. The GMS, which is a concept originated by Neelin and Held (1987), represents the efficiency of MSE export by convection and associated large-scale circulations. Raymond et al. (2009) defines a relevant quantity called normalized GMS (NGMS), which is a ratio of column MSE (or moist entropy) advection to intensity of the convection. Although different authors have used
slightly different definitions of the NGMS (e.g., Fuchs and Raymond 2007; Raymond and Fuchs 2009; Raymond et al. 2009; Sugiyama 2009; Andersen and Kuang 2011), the physical implications behind those definitions are consistent in such a way that the NGMS represents efficiency of export of some intensive quantity conserved in moist adiabatic processes per unit intensity of the convection (Raymond et al. 2009). We employ one version of the NGMS defined as

\[
\Gamma = \frac{\langle \vec{v} \cdot \nabla h \rangle + \langle \omega \frac{\partial h}{\partial p} \rangle}{\langle \vec{v} \cdot \nabla s \rangle + \langle \omega \frac{\partial s}{\partial p} \rangle},
\]

(2)

where \( h \) and \( s \) represent MSE and DSE, respectively. Since in the tropics, horizontal temperature gradients are negligible (weak temperature gradient; Sobel and Bretherton 2000), neglecting the horizontal DSE advection in the denominator yields

\[
\Gamma = \frac{\langle \vec{v} \cdot \nabla h \rangle + \langle \omega \frac{\partial h}{\partial p} \rangle}{\langle \omega \frac{\partial s}{\partial p} \rangle}.
\]

(3)

Equation 3 can be separated into horizontal and vertical components as

\[
\Gamma = \Gamma_h + \Gamma_v
\]

(4)

where

\[
\Gamma_h = \frac{\langle \vec{v} \cdot \nabla h \rangle}{\langle \omega \frac{\partial s}{\partial p} \rangle},
\]

\[
\Gamma_v = \frac{\langle \omega \frac{\partial h}{\partial p} \rangle}{\langle \omega \frac{\partial s}{\partial p} \rangle}.
\]

In some NGMS studies, the vertical component of the NGMS \( \Gamma_v \) is simply called NGMS (or GMS) (e.g., Sugiyama 2009; Kuang 2010; Andersen and Kuang 2011; Sobel and Maloney 2012) while in the others, the horizontal component \( \Gamma_h \) is explicitly defined (e.g., Raymond and Fuchs 2009; Raymond et al. 2009; Benedict et al. 2014; Hannah and Maloney 2014; Sobel et al. 2014). \( \Gamma_v \) has been used in various ways such as a diagnostic quantity in general circulation models (e.g., Frierson 2007; Hannah and Maloney 2011, 2014; Benedict et al. 2014), in observational data (e.g., Yu et al. 1998; Sobel et al. 2014)\(^2\), as an output quantity.

\(^2\)In Yu et al. (1998), the computed quantity was GMS, and not normalized one.
of a MJO toy-model (e.g., Raymond and Fuchs 2009), and as an input parameter of a MJO
toy-model (e.g., Sugiyama 2009; Sobel and Maloney 2012, 2013). As Hannah and Maloney
(2011) and Masunaga and L’Ecuyer (2014) pointed out, values of $\Gamma_v$ generally fluctuate
in convective life-cycles primarily due to variations of vertical velocity profiles (as seen in
Fig. 3). Nevertheless, when used as an input parameter of a toy-model, $\Gamma_v$ is assumed to
be a constant in the convective life-cycle (e.g., Sugiyama 2009; Sobel and Maloney 2012).
Furthermore, time-dependent fluctuations of the NGMS are also neglected when the NGMS
is computed based on scatter plots between the numerator and denominator of the NGMS,
which is one of the most general methods to compute the NGMS.

When considering the NGMS on different time scales in data, we have to be careful
about its interpretation. First of all, we can define a mean NGMS, in which we average
the numerator and the denominator of $\Gamma$ before taking the ratio. This is in keeping with
the spirit of the definition. We can also define an anomalous NGMS, in which perturbations
from the means of numerator and denominator are taken and the ratio of these perturbations
is computed. Similarly, we can define a total NGMS. It can be easily shown that the total
NGMS is a constant if and only if the mean NGMS is equal to the anomalous NGMS. In
many of previous studies, the total NGMS has been assumed to be constant. In such cases,
one doesn’t have to worry about the differences between the mean and anomalous NGMS.
But when considering the total NGMS as a time-dependent variable, one should clarify which
kinds of NGMS are being used, mean, anomalous, or total NGMS.

Furthermore, we can generalize the idea of the decomposition of the NGMS from an
aspect of Fourier transformation. By taking Fourier decomposition, Eq. 1 can be separated
into

$$\frac{\partial \langle h \rangle_i}{\partial t} = -\langle \vec{v} \cdot \nabla h \rangle_i - \langle \omega \frac{\partial h}{\partial p} \rangle_i + \langle Q_R \rangle_i + SF_i$$

(5)

The phrase "total NGMS" is often used to refer to the combination of $\Gamma_h$ and $\Gamma_v$. In this study, we
use the phrase "total NGMS" to refer to the combination of anomaly and mean state. $\Gamma_h$ plus $\Gamma_v$ is simply
called NGMS or $\Gamma$ in this paper.
where subscripts represent a specific range of frequencies. For instance, $i = 0$ can be defined as the mean state, and $i = ISO$ can be defined so that Eq. 5 represents intra-seasonal oscillations as in Maloney (2009). Therefore, we can define the NGMS on different time scales as

$$
\Gamma_i = \langle \vec{v} \cdot \nabla h \rangle_i + \langle \omega \frac{\partial h}{\partial p} \rangle_i \langle \omega \frac{\partial s}{\partial p} \rangle_i.
$$

The horizontal and vertical components on different time scales can be defined similarly to Eq. 4.

Interpretations of the sign of the NGMS also require some attentions. When dealing with band-pass filtered variability, the denominator of Eq. 6 represents anomalous quantities which can be both positive and negative. With a positive denominator (this is a usual case when convection is active), positive/negative NGMS corresponds to export/import of the MSE. But, when the denominator is negative (or when convection is inactive), the interpretation must be reversed; that is, a positive/negative value corresponds to import/export of the MSE.

b. NGMS during TOGA COARE

We estimated the time-dependent NGMS on the four different time scales using Eq. 6. Figure 5 shows the lag regression slopes of horizontal (blue), vertical (red), and combined (green) column-integrated MSE advection as a function of lag regression slopes of column-integrated vertical DSE advection on the different time scales. The elliptic shapes represent life-cycles of convection in which each life-cycle starts from a filled circle, going around counterclockwise, and terminates at a filled square. $\Gamma_h$, $\Gamma_v$, and $\Gamma$ at different convective phases can be estimated by computing the slopes of the lines which are drawn from the origin to the periphery of the elliptic shapes. For instance, on $\sim$2 day scale, $\Gamma_v$ starts with a positive value ($\sim$0.2) which becomes larger and goes infinity (this corresponds to the singularity of the NGMS). After passing through the singular point, it becomes negative.
which grows into a positive value and reaches about 0.2 again at the peak of the convection. After the convective peak, $\Gamma_v$ increases and becomes infinity again at the singular point, followed by negative values.

One conclusion we can draw from Fig. 5 is that the NGMS and all the components are not constant values on all the time scales, but they vary along the convective life-cycle. But we can find that as the time scale gets longer the vertical NGMS $\Gamma_v$ converges to a constant value around 0.2, which is the slope of the major axis of the elliptic shape. On the MJO time scale, the elliptic shape of the vertical MSE advection becomes very close to a linear shape (i.e., constant $\Gamma_v$) with the minor axis collapsed. This more-constant $\Gamma_v$ is related to the fact that the column-integrated vertical MSE advection becomes closer to 180 degree out of phase (negatively in phase) with the precipitation as the time scale gets longer. This indicates that on longer time scales, the column-integrated vertical MSE advection is more linearly correlated to the precipitation. This result might support one of the popular usages of $\Gamma_v$ in a MJO toy-model in which $\Gamma_v$ is assumed be a time-independent quantity (e.g., Sugiyama 2009; Sobel and Maloney 2012).

Compared with the vertical advection, the horizontal advection doesn’t have a consistent pattern among the different time scales. On $\sim$2 day scale, the major axis of the ellipse of the horizontal advection has a positive slope while on $\sim$5 day scale the slope is almost zero. In contrast, on $\sim$10 day scales, it has a negative slope. On the MJO scale, its slope is slightly positive, but the values of $\Gamma_h$ vary significantly during the convective life-cycle. As a result, the NGMS $\Gamma$ (combination of $\Gamma_h$ and $\Gamma_v$) also varies significantly during the convective life-cycles on all the time scales. It should also be noted that the elliptic patterns of $\Gamma$ are more similar to those of $\Gamma_v$ than those of $\Gamma_h$ on all the time scales except for the MJO time scale.
5. Discussion

a. Omega profiles and $\Gamma_v$

Most of the variations in $\langle \omega \partial h / \partial p \rangle$ are explained by the variations of the omega profiles (94% of the total variance in the TOGA COARE data), and the variations of the MSE profiles play a small role. We can use the assumption that omega profiles can be approximated by two dominant modal structures to reason about the importance of each mode for the column MSE budget. We assume

$$\omega(t, p) \approx o_1(t) \Omega_1(p) + o_2(t) \Omega_2(p)$$  \hspace{1cm} (7)

where $\Omega_1$ and $\Omega_2$ are often called first and second baroclinic modes, respectively, and $o_1$ and $o_2$ represent the time-dependent amplitudes of those modes. These could be any two modes which do a good job of describing the variability in vertical motion profiles, like those that come from a principle component analysis of vertical motion profiles. In the TOGA COARE data, the first mode of a principle component analysis (PCA) explains 71% of the variance, and the second mode explains 21% of the total variances of the omega profiles.

If we neglect the variations of the MSE profiles, we can represent $\Gamma_v$ as

$$\Gamma_v \approx \frac{o_1 \langle \Omega_1 \partial h / \partial p \rangle + o_2 \langle \Omega_2 \partial h / \partial p \rangle}{o_1 \langle \Omega_1 \partial \bar{s} / \partial p \rangle + o_2 \langle \Omega_2 \partial \bar{s} / \partial p \rangle}$$  \hspace{1cm} (8)

where the bars represent the time averages.

In general, the MSE and DSE profiles, $\Omega_1$ and $\Omega_2$, if chosen via PCA, have the structures as shown in the schematic figure, Fig. 6. In the first baroclinic system, convergence happens in the lower troposphere where the DSE is poor and divergence happens in the upper troposphere where the DSE is rich. Hence, in this system, strong net export of DSE happens (i.e., $\langle \Omega_1 \partial \bar{s} / \partial p \rangle$ is positive and large). In contrast, in the second baroclinic system, convergence happens both in the lower and upper troposphere where the DSE is poor and rich, respectively, and divergence happens in the middle troposphere where the DSE is moderate. As a result, the upper tropospheric net import of DSE is canceled out by the lower tropospheric
net export of DSE, causing small value of \( \langle \Omega_2 \partial \tilde{s} / \partial p \rangle \). Consequently, the value of \( \langle \Omega_1 \partial \tilde{s} / \partial p \rangle \) is much larger than \( \langle \Omega_2 \partial \tilde{s} / \partial p \rangle \). Neglecting \( \langle \Omega_2 \partial \tilde{s} / \partial p \rangle \) in Eq. 8 yields

\[
\Gamma_v \approx \frac{\langle \Omega_1 \frac{\partial \bar{h}}{\partial p} \rangle}{\langle \Omega_1 \frac{\partial \bar{h}}{\partial p} \rangle} + \frac{a_2}{a_1} \frac{\langle \Omega_2 \frac{\partial \bar{h}}{\partial p} \rangle}{\langle \Omega_1 \frac{\partial \bar{h}}{\partial p} \rangle}.
\]

(9)

This equation shows that for this set of assumptions, time-dependent fluctuations \( \Gamma_v \) are due to the second term in the rhs of Eq. 9, which is the ratio of the amplitude of the second mode to that of the first mode times the ratio of the gross moist stability due to the second mode to the gross dry stability (the denominator of \( \Gamma_v \); Yu et al. 1998) due to the first mode. In general, \( \langle \Omega_2 \partial \bar{h} / \partial p \rangle \) is negative and large while \( \langle \Omega_1 \partial \bar{h} / \partial p \rangle \) is positive and small (based on Fig. 6 and similar arguments to those for the gross dry stability \( \langle \Omega_1 \partial \tilde{s} / \partial p \rangle \) and \( \langle \Omega_2 \partial \tilde{s} / \partial p \rangle \)).

Thus, for this set of assumptions, the second term in the rhs of Eq. 9 is responsible for negative \( \Gamma_v \) in the early stage of the convection, as pointed out by Hannah and Maloney (2011) and Masunaga and L’Ecuyer (2014). This term is also responsible for the nonlinearity of the vertical MSE advection with respect to the convection, making the elliptic trajectories in Fig. 5. If this time-dependent term disappears, \( \Gamma_v \) given by Eq. 9 is the homomorphism of the GMS given by Neelin and Held (1987).

In Fig. 3, we showed that as the time scale gets longer, the tilting structure of the omega profile becomes less prominent. This disappearance of the tilt is likely due to smaller contributions of the second baroclinic mode on longer time scales compared to those on shorter time scales. This indicates that the second term in the rhs of Eq. 9 becomes smaller as the time scale gets longer, making \( \Gamma_v \) a more time-independent quantity. On shorter time scales where the second baroclinic mode is prominent, in contrast, the time-dependent term in Eq. 9 is robust, hence \( \Gamma_v \) on those time scales varies significantly in the convective life-cycles.

Some studies have argued for an important role of shallow convection in the convective variability including the MJO in which shallow convection enhances moisture import via enhanced surface convergence, and thus amplifies the convective system (e.g., Wu 2003; Kikuchi and Takayabu 2004). In our results, although it was less significant than the deep
convective profile, a shallow convective phase can be observed even on the MJO time scale. That shallow convection could play a role in the MJO dynamics.

Interestingly, the elliptic trajectories shown in Fig. 5 have been already pointed out by Masunaga and L’Ecuyer (2014), who investigated the MSE budgets and computed the time-evolution of the NGMS on short time scales using the satellite data sets. There are a few notable differences between our analysis and their study. First, they used a different NGMS definition, which is a ratio of MSE advection to moisture advection instead of DSE advection. Therefore, their NGMS plot is a mirror image of our NGMS plot with respect to the x-axis (see Fig. 13 in Masunaga and L’Ecuyer (2014)). Second, they computed the total NGMS including the background state instead of the anomalous NGMS which we computed. Thus the center of the elliptic shape is shifted to the right and downward. The composite methodology is also different from our study. Nevertheless, their study has drawn a similar conclusion about the NGMS variability to ours. That is, the first/second baroclinic modes explain the larger (along the major axis)/smaller (along the minor axis) variability of the elliptic trajectory.

b. How to compute NGMS

The values of estimated NGMS depend on the method of the computation. In section 4, we showed the NGMS as a time-dependent variable. But in some recent NGMS studies, NGMS is computed based on a scatter plot of MSE advection as a function of DSE advection (e.g., Raymond and Fuchs 2009). In such a case, time-dependent fluctuations are not taken into account.

If we estimate the NGMS following that method, then the values of the NGMS, the horizontal and vertical components correspond to the slopes of the major axes of the elliptic trajectories in Fig. 5. The values of those slopes (\( \bar{\Gamma}, \bar{\Gamma}_h, \) and \( \bar{\Gamma}_v \)) are summarized in Table 1. As discussed above, \( \bar{\Gamma}_h \) varies significantly among the time scales. Consequently, \( \bar{\Gamma} \) which is the combination of \( \bar{\Gamma}_h \) and \( \bar{\Gamma}_v \) also varies among the different time scales. Although smaller
than the variations of $\tilde{\Gamma}$ and $\tilde{\Gamma}_h$, there are variations of $\tilde{\Gamma}_v$ among the time scales, too. These might be due to the variations of the shapes of $\Omega_1$ among the different time scales, which could be caused by errors due to the small number of the independent samples.

c. Tilt in other work

Mapes et al. (2006, hereafter M06) proposed the "Stretched Building Block" hypothesis that "individual cloud systems in different phases of a large-scale wave have different durations of shallow convective, deep convective, and stratiform anvil stages in their life cycles.” This hypothesis was proposed to explain the apparent multi-scale similarities of the vertical structures between the mesoscale convective systems, convectively coupled equatorial waves, and the MJO. The systematic steepening of the leading edge slopes in the omega profiles shown in Fig. 3 suggest that omega may not have as much multi-scale similarity as M06 suggested, especially on the MJO time scale.

The wind divergence field on the MJO time scale in our result (Fig. 3h) resembles that in M06 (the second panel of Fig. 8 therein), both of which contain a small amount of tilt. However, that tilt is, as shown in section 3b, too small to claim the multi-scale similarity of the omega profiles, especially on the MJO time scale. In contrast, a significant multi-scale similarity is observed in the specific humidity field. Figure 7 shows the time-height structures of specific humidity on the different time scales and there is significant tilt on all time scales, unlike in vertical motion. Our figure is consistent with Fig. 7 in M06, which is given as evidence for the vertical tilt in clouds on longer time scales. Hence, we conclude that tilt in the moisture field is more robust than that in the omega field on the MJO time scale.

Previous work is also suggestive of more tilt in diabatic heating than we are finding, during the TOGA COARE MJO. Especially, our results can be compared with Fig. 9 in Lin et al. (2004, hereafter L04) and Fig. 12 in Kiladis et al. (2005, hereafter K05), in which the TOGA COARE data set was analyzed in a similar lag-regression method to ours. These studies examined diabatic heating (or Q1), which has a very similar structure to omega.
The major difference in results between these studies and ours is found in the tailing edges of the event, where the L04 and K05 figures have more tilt. In Figs. 3 and 4, we show our lag-regressed plot resembles the raw structure of the MJO with a simple time filter. We believe that the relevant difference in methodology between their work and ours is that both of the other studies used spatial filters in addition to time filters to obtain their index time-series. Personal communication with Kiladis and Haertel confirmed that spatial filtering was used in their analysis and that the difference of time versus time-space filters makes non-negligible differences in the diabatic heating structures.

**d. Sensitivity of choice of filter**

Finally we will briefly discuss sensitivity of the choice of filters. Figure 8 illustrates the response functions of the >1.5 day low-pass Lanczos filter and daily running mean filter. This figure shows that by using the running mean filter, about 60% of the signals on 2 day scale are lost due to the shallow slope of the response function. Even at 4 day period which corresponds to the time scale of some of the Kelvin waves, about 20% of the signals are lost. This indicates that for examining high frequency variability such as inertia gravity waves or Kelvin waves, the Lanczos filter with a steeper slope of the response function is more appropriate than the running mean filter.

**6. Conclusions**

We have examined the column-integrated moist static energy (MSE) budget during the TOGA COARE field campaign, using sounding data, and filtering the data into various frequencies of variability with ∼2 day, ∼5 day, ∼10 day, and >20 day periodicity. In the deep tropics, fluctuations of the column MSE are primarily due to variations of column-integrated water vapor which are tightly connected with precipitation anomalies. Therefore, investigating the mechanisms of recharge and discharge of the column MSE leads us to a better
understanding regarding the convective amplification and decay. Our analysis highlights the importance of the investigation of the column MSE on different time scales. We found that each budget term of the column MSE behaves in significantly different ways on the different time scales. As a result, dominant processes in the MSE recharge and discharge differ among the time scales. Some notable results are summarized as follows:

i. On all the time scales except for the MJO time scale, the vertical MSE advection, $-\langle \omega \partial h / \partial p \rangle$, is the most dominant process with the greatest magnitude of variations in the MSE recharge-discharge mechanism.

ii. On the shorter time scales ($\sim 2$ day and $\sim 5$ day scales), the vertical MSE advection accounts for most of the MSE recharge and discharge, and the other terms cancel out each other so that the tendency of the column MSE $\langle \partial h / \partial t \rangle$ is primarily explained by the vertical MSE advection.

iii. As the time scale gets longer, the relative importance of the other terms than the vertical advection becomes greater. Especially on the MJO time scale, all the budget terms (horizontal advection, $-\langle \vec{v} \cdot \nabla h \rangle$, vertical advection, $-\langle \omega \partial h / \partial p \rangle$, radiative heating $\langle Q_R \rangle$, and surface fluxes $SF$) have nearly the same magnitude of variations.

iv. The horizontal advection behaves in significantly different ways among the different time scales.

v. The amplitude of the horizontal advection becomes greater as the time scale gets longer, indicating that the horizontal advection plays a more important role in the MSE recharge-discharge mechanism on longer time scales than shorter time scales.

vi. The radiative heating is approximately in phase with the precipitation (or the precipitation leads slightly) while the surface fluxes lag the precipitation except for $\sim 10$ day scale on which both the radiative heating and surface fluxes are approximately in phase with the precipitation.
vii. On the shorter time scales, the MSE export via vertical advection (i.e., positive $\langle \omega \partial h / \partial p \rangle$) lags the precipitation peak. As the time scale gets longer, however, the MSE export becomes more in phase with the precipitation.

The last bullet of the summary above, more in-phase MSE export via vertical advection, is primarily explained by variations in the omega profile. The tilt of the profile at the leading edge of the convection gets steeper as the time scale gets longer. This implies that the second baroclinic structure of the omega profile becomes less robust in the early stage of the convection. On the MJO time scale, the leading edge tilt becomes very steep, and the overall omega structure becomes closer to the first baroclinic mode. Consequently, the vertical component of the normalized gross moist stability (NGMS) becomes more a constant quantity which is nearly independent of the convective life-cycle. In contrast, on the shorter time scales where a second baroclinic mode is prominent, the vertical NGMS has large time-dependency, thus the values of the vertical NGMS vary significantly along the convective life-cycle. The horizontal component of the NGMS doesn’t have a consistent pattern among the different time scales since the horizontal MSE advection behaves in significantly different ways on the different time scales.

Furthermore, our results shown in Fig. 2d, the MSE budgets in the MJO event, reinforce the view of the MJO dynamics which has been emerging from recent MJO studies (e.g., Kim et al. 2014; Sobel et al. 2014) in the following ways: 1) The radiative heating and surface fluxes destabilize the MJO disturbance by amplifying and maintaining MSE anomalies. 2) The vertical advection stabilizes the disturbance by exporting the MSE and canceling the effects of the radiative heating and surface fluxes. 3) The horizontal advection plays a significant role in the eastward propagation by providing moistening ahead (in the negative lags, or to the east of), drying behind (in the positive lags, or to the west of) the active phase. Although there are differences between the different MJO events, our results in general show significant commonalities with those view points.
Finally, we should acknowledge again that we are uncertain about whether or not the
results shown for the longer time scale variability (∼10 day and the MJO time scales)
represent statistically significant patterns because of the lack of the degrees of freedom.
Our results for the MJO timescale are broadly consistent with published work on MSE
budgets observed during the DYNAMO field campaign by Sobel et al. (2014), though we
find the vertical NGMS less variable over an MJO life-cycle, possibly due to our use of the
Lanczos filter rather than a running mean. For more accurate and solid conclusions, we need
to investigate more data sets such as ERA-Interim and TRMM which contain much longer
time-series than the TOGA COARE data. We would also like to repeat our analysis using
DYNAMO data in future work.

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<th>$\sim5$ day scale</th>
<th>$\sim10$ day scale</th>
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<td>0.25</td>
<td>0.29</td>
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