

OTREC, PREDICT and EC model to study convection

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Key points:

- Both boundary layer and free tropospheric characteristics are important for development of convection.
- Field project and modeling data agree on the interplay of thermodynamic parameters.

Abstract

Two data sets (OTREC and PREDICT) and the European Center operational model are used to study convection.

Vertically integrated moisture convergence is taken to be a proxy for convection. The interplay between the thermodynamic parameters is examined for weak and strong convection. It is found that for strong convection saturation fraction anti-correlates with instability index and DCIN in a similar manner for all analyzed regions of PREDICT and OTREC.

The European Center model is used to look at the composite diurnal cycles of strong and weak convection. As convection increases, saturation fraction increases while instability index and DCIN decrease as in OTREC and PREDICT. The boundary layer and free troposphere thus both play an important role in convection development.

Plain Language Summary

Convection in the tropics is the main mechanism that brings bad weather, storms and hurricanes there. Understanding how convection develops and what it depends on is a question that we have been trying to answer for many decades. The characteristics of the environment can help in this task. The data from two field projects, OTREC and PREDICT, as well as from the European Center model are analyzed here to show what characteristics of the environment are important for convection. This can help us in improving the weather models and forecasts that are very much needed in the tropics.

1. Introduction

Mesoscale convective systems (MCS) are responsible for most rain in the tropics (Houze, 1989). They are ensembles of convection that are often idealized as entraining plumes of 1 - 5 km diameter that go through their life cycle almost independently of one another. If the convection grows in time that implies that the number of plumes is getting larger and convection covers a larger area, while when convection decays, their number is getting smaller (Arakawa and Shubert 1974; Zipser, 1969, 1977; Houze, 1989, 2004; etc.; Ed Zipser, personal communication).

Another school of thought tries to study the broader properties of MCSs and most importantly their interaction with the environment (Raymond et al., 2011). This approach does not look at the internal structure of the MCSs, but how the environment governs their overall characteristics. The goal is to identify the thermodynamic properties of the environment that govern MCSs. If one gets those right, it would be possible to correctly parametrize convection (Raymond and Fuchs-Stone, 2020).

What properties of the environment guide MCSs? Is it the boundary layer, the free troposphere or both? There are many papers that suggest that the boundary layer is responsible for the intertropical convergence zone, ITCZ (Riehl et al., 1951; Lindzen and Nigam, 1987; Battisti et al., 1999; Tomas et al., 1999; Stevens et al., 2002). Back and Bretherton (2009) postulate two types of convection, one that is generated by the boundary layer convergence and another that is generated by the free troposphere noting that the boundary layer convergence is of more importance in east and central

Pacific ITCZ. Raymond (2017) advocates for the importance of thermodynamic factors in this region, mainly in the free troposphere.

Based on Fuchs-Stone et al. (2020) and Raymond and Fuchs (2020), we hypothesize that saturation fraction, instability index and DCIN play an important role in development of convection.

The saturation fraction is defined as precipitable water divided by the saturated precipitable water and it is a measure for column relative humidity.

The instability index is defined as the difference between the saturated moist entropies averaged over the altitude ranges of 1-3 km and 5-7 km. The instability index is a measure of low to mid-tropospheric moist convective instability. Lower, but still positive, values are associated with higher saturation fraction and more rainfall (Raymond et al., 2014, Gjorgjievska and Raymond, 2014, Raymond et al., 2015, Sentic et al., 2015, Raymond and Flores, 2016, Singh et al., 2019, Raymond and Kilroy, 2019). We call this association “moisture quasi-equilibrium”.

Deep convective inhibition (DCIN) is defined as the difference between the average of the saturated moist entropy in the 1.5 - 2 km layer and the boundary layer moist entropy averaged over the 0 - 1 km layer.

In this paper we choose vertically integrated moisture convergence, represented in energy terms, to be to be a proxy for convection. It is defined as:

$$MC = -L \int \nabla \cdot (\rho \mathbf{v} r)$$

where ρ is the density, \mathbf{v} is the horizontal velocity, r is the mixing ratio and L is the latent heat of condensation. It should be noted that there are other degrees of freedom that might be active in determining what governs convection. We choose vertically integrated moisture convergence because it is a good parameter when estimating rainfall. Based on moisture convergence, we classify convection into strong and weak categories. We then look at the interplay between the thermodynamic parameters and convection.

We use the dropsonde data sets from two field projects: OTREC (Organization of Tropical East Pacific Convection) which took place in 2019 and PREDICT (Pre-Depression Investigation of Cloud-Systems in the Tropics) which took place in 2010.

OTREC studied all phases of convection over the East Pacific and Southwest Caribbean from Liberia, Costa Rica using the NSF/NCAR Gulfstream V aircraft (Fuchs-Stone et al., 2020). Twenty-two research missions were flown and 648 dropsondes were successfully deployed in a grid of three regions of diverse environments. The research flight area in the Eastern Pacific ITCZ, B2, covers the region with strong meridional sea surface temperature (SST) gradients, while B1b, the Caribbean region, covers the area with uniform and moderately high SSTs. The B1a box is located just off the Pacific coast of Colombia and it covers the environment of lower SSTs in the proximity of land.

PREDICT studied systems with potential to develop into tropical cyclones in the Western Atlantic and Caribbean using the NSF/NCAR Gulfstream V aircraft (Montgomery et al., 2012). Twenty-six research missions were flown and 547 dropsondes were deployed. The PREDICT data covers a vast region with high SSTs and higher latitudes than OTREC.

94

95 Three-dimensional variational analysis (3DVAR) of dropsonde data is performed to obtain regular
 96 grids of data 0.5 by 0.5 degree for OTREC and 1 by 1 degree for PREDICT (Lopez and Raymond,
 97 2011; Raymond and Lopez, 2011; Raymond et al., 2011; Montgomery et al., 2012; Gjorgjievska
 98 and Raymond, 2014; Juracic and Raymond, 2016; Fuchs-Stone et al., 2020; Raymond and Fuchs-
 99 Stone, 2020). The data set for dropsondes used in this paper is NCAR/EOL AVAPS Dropsonde
 100 Quality Controlled Data Version 1.0, Voemel, H. (2019). The grids for OTREC and PREDICT are
 101 different to avoid sample size artifact.

102 Finally, we look at the time evolution, i.e. composite diurnal cycle of the thermodynamic
 103 parameters in strong and weak convection using the operational European Centre for Medium-
 104 Range Weather Forecasts (ECMWF) model.

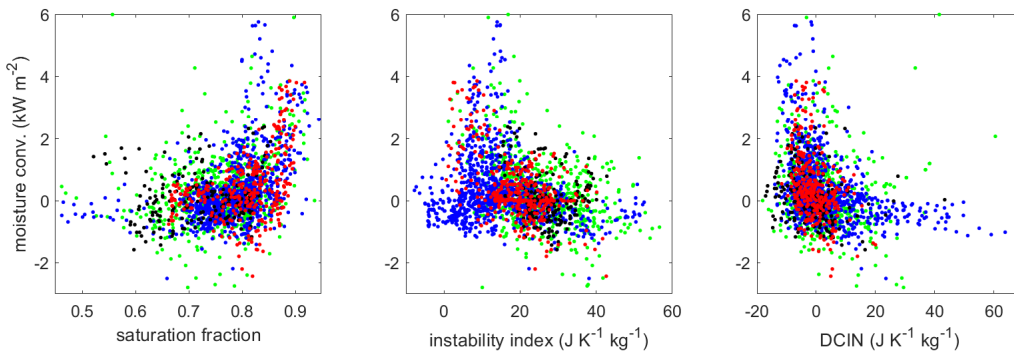
105 Section 2 presents the results and their interpretation, while conclusions are given in section 3.

106 2. Results

107

108 2.1 The interplay between convection and thermodynamic parameters

109 Figure 1 shows how our thermodynamic parameters saturation fraction, instability index and DCIN
 110 correlate with moisture convergence, i.e. convection. Each dot for OTREC data set represents the
 111 average in half a degree by half a degree box obtained from the 3DVAR, one by one degree average
 112 for PREDICT data. PREDICT data is shown in green, the data for the Eastern Pacific ITCZ box
 113 B2 in blue, the Pacific box off the coast of Colombia, B1a, is shown in red and the data for the
 114 Caribbean B1b box is shown in black. We see that higher values of saturation fraction lead to more
 115 convection. The opposite is true for the instability index where lower values correspond to more
 116 convection. Negative and small positive values of DCIN lead to more convection as well
 117 (Raymond et al., 2003).

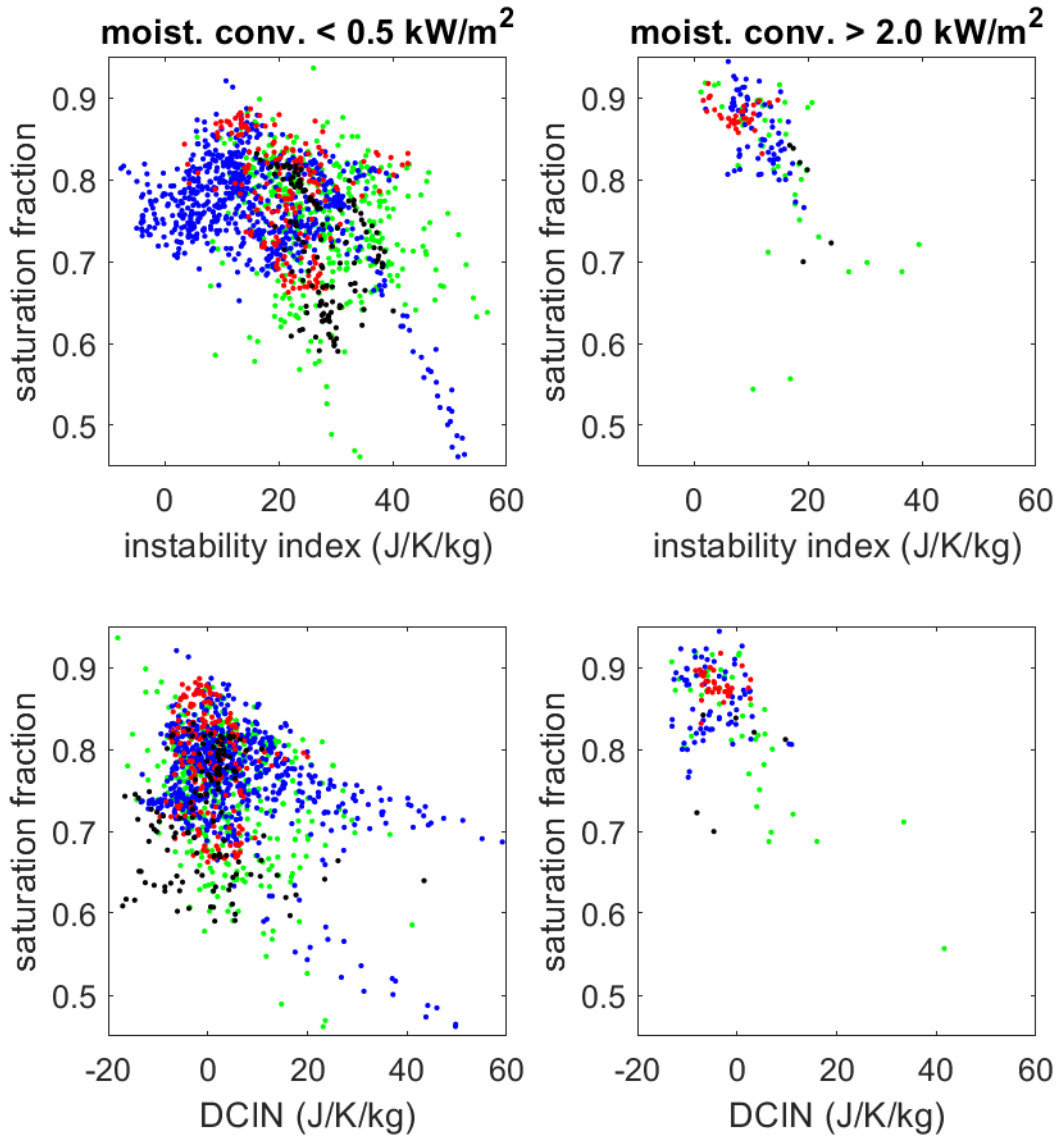


118

119 **Figure 1:** Scatter plot of moisture convergence and saturation fraction, instability index and DCIN.
 120 Green dots are PREDICT data, blue Eastern Pacific ITCZ, red Colombian box and black Caribbean
 121 box.

122 To decipher Figure 1, that includes all cases from null to strong convection, we now set out to look
 123 at the relationships between our thermodynamic parameters saturation fraction, instability index
 124 and DCIN for strong and weak convection. We define convection by the strength of vertically

125 integrated moisture convergence. If moisture convergence is greater than 2 kWm^{-2} we define it
 126 as strong convection and if it is less than 0.5 kWm^{-2} we define it as weak convection.



127
 128 **Figure 2:** Scatter plots between saturation fraction and instability index (top panel) and saturation
 129 fraction and DCIN (bottom panel) for weak convection (left) and strong convection (right).

130 Figure 2 shows scatter plots between saturation fraction and instability index and saturation
 131 fraction and DCIN for strong convection and weak convection for PREDICT and OTREC data.
 132 For strong convective cases the saturation fraction and instability index show a clear anti-
 133 correlation as per moisture quasi-equilibrium theory. The scatter plot between saturation fraction
 134 and DCIN shows that negative or very small DCIN is required for strong convection. The scatter
 135 plot for weak convection is significantly more scattered (including some outliers in B2 and
 136 PREDICT regions). The saturation fraction can be lower than expected in this case. This is because

the environment in weak convection is too dry and moisture quasi-equilibrium theory does not hold. DCIN turns towards higher positive values.

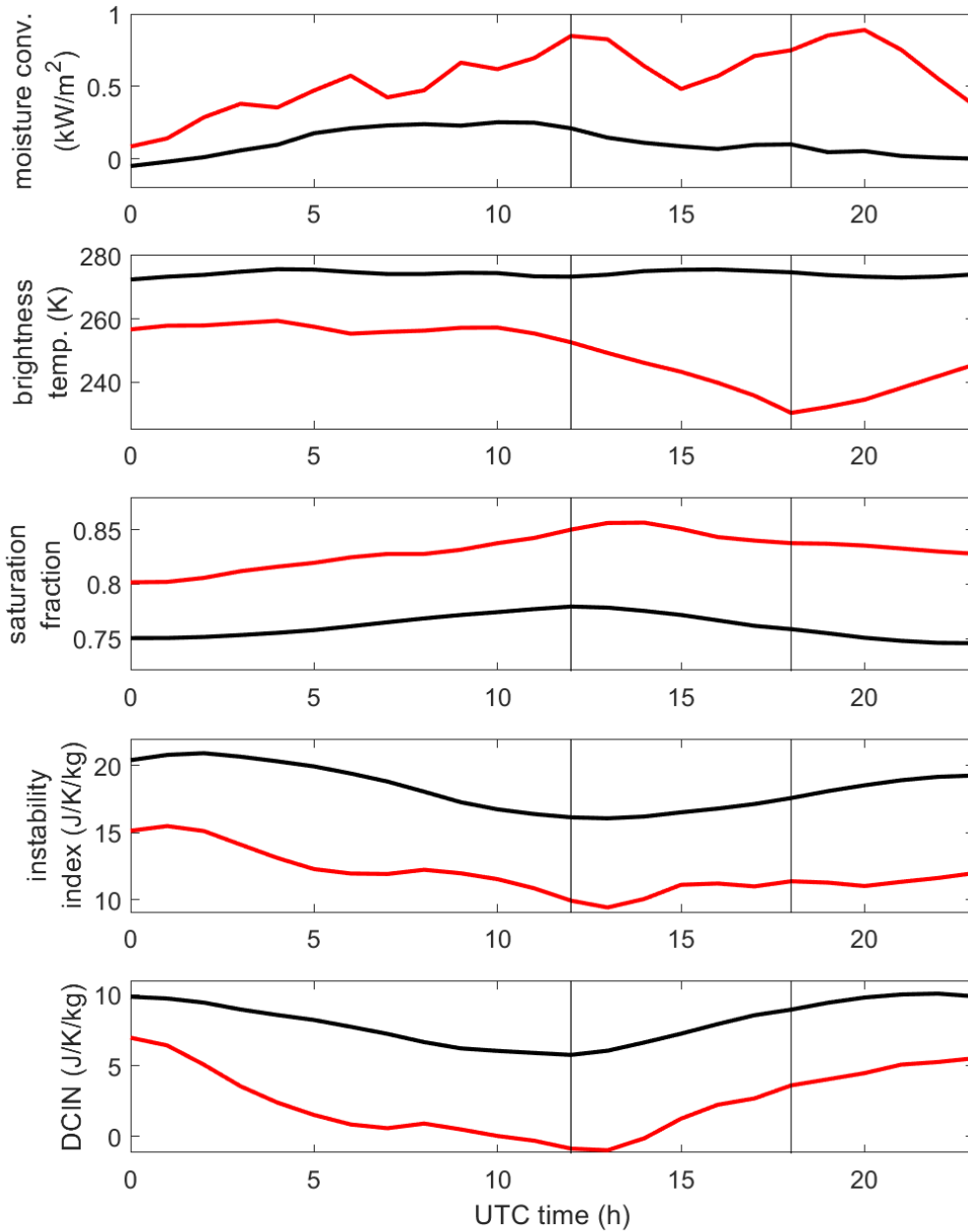
Figures 1 – 2 tell an interesting story. Strong convection seems to be similar in all regions irrespective of sea surface temperature (SST) (high in PREDICT, lower in B1b and even lower in B2 and B1a), whether we are close to land (B1a box) or not. Once there is significant convection, there is no doubt that moisture convergence depends on high saturation fraction, low instability index and low DCIN. Furthermore, saturation fraction anti-correlates with instability index and DCIN in a similar manner. When convection is weak and the environment is dry, the time scale for convection to adjust the moisture to the equilibrium value for a given value of instability index is long and this anti-correlation is not necessarily observed. DCIN takes up higher values indicating inversion.

2.2 Composite diurnal cycle

To look at the time evolution of convection for OTREC we look at the composite diurnal cycles from the operational EC model for days with OTREC observations. We first use the 3DVAR moisture convergence data to define strong convective and weak convective cases. We then average the ECMWF operational model data in boxes with the same longitude and latitude for all times of the same day. Then we average all the hourly values. The same method is applied to the Geostationary Operational Environmental Satellite (GOES-16) brightness temperature.

Figure 3 shows brightness temperature from the GOES-16 and moisture convergence, saturation fraction, instability index and DCIN from the EC model averaged over all strong convective cases (red line) and weak convective cases (black line) as a function of time. The vertical lines bound the time of a day when OTREC dropsondes were deployed and assimilated into the EC model (12 to 18 UTC).

We see that the convection represented by moisture convergence peaks at 12 hours UTC, 0600 local time. This agrees with our observations while in the field. On average convection was developing in the very early morning hours, becoming stratiform around 10 am local time, 16 UTC. The brightness temperature from satellite observations GOES-16 for the same cases has a minimum that lags the maximum of the moisture convergence from the EC model. This is due to a well-known lag between convection and resulting stratiform cloudiness as the upper troposphere needs time to become optically thick (Bechtold et al., 2014). This gives us confidence in the moisture convergence maximum at 12 UTC from the EC model. The second maximum in moisture convergence that occurs at 20 UTC appears to be related to the overestimation of the diurnal convection over nearby land. Note that spurious moisture convergence in EC at 20 UTC is not supported by saturation fraction, instability index and DCIN values. The brightness temperature does not support it either. Extensive research and comparison between satellite loops and the EC model at 20 UTC show that frequently EC develops nonexistent convection in the B2 box. This is known to the ECMWF and it was one of the reasons for their support of OTREC project and the location of B2 box.



175

176 **Figure 3:** Moisture convergence from EC model, GOES-16 brightness temperature, saturation
 177 fraction, instability index and DCIN from EC model averaged over strong (red line) and weak
 178 (black line) convective cases from OTREC.

179 Prior to the peak in the moisture convergence at 12 UTC, we see that the saturation fraction
 180 increases. During that same time period the instability index and DCIN decrease. They reach their
 181 maximum (saturation fraction) and minima (instability index and DCIN) very close to the

maximum in moisture convergence. Our confidence in time evolution of saturation fraction, instability index and DCIN from the EC operational model is high as we have compared the EC model runs with and without dropsondes assimilated (not shown here). The agreement between the two is very good.

The weak time evolution represented by black line shows in contrast to convective line (in red) that moisture convergence is significantly less as is the saturation fraction. Instability index and DCIN have higher values.

3. Conclusions

The goal of this study is to study convection using data from two field projects, OTREC and PREDICT, and the ECMWF model.

Some of the questions that we try to answer are: Does physics of convection change with different SSTs and is it different if there is proximity to land? How does convection change when we go from strong to weak convection? Does the dependence on saturation fraction, instability index and DCIN change?

Vertically integrated moisture convergence is taken as a proxy for convection. This assumption is not ideal, but it is chosen to represent rainfall and therefore convection. Convection is classified into strong and weak categories. We look at the interplay of convection and thermodynamic parameters saturation fraction, instability index and DCIN for different phases of convection.

Strong convection looks very similar regardless to the environment, SST, proximity to land, different latitude etc. Saturation fraction shows anti-correlation with instability index and DCIN for every region that we have looked at. The fact that negative or small positive DCIN is needed for convection comes as no surprise (Raymond et al., 2003), while the instability index impact on convection has been less researched. We conclude that both, the boundary layer (DCIN) and free troposphere (saturation fraction and instability index) are very important for development of convection. As we move towards weak convection, the story becomes more complicated. The above correlations are less pronounced and more scattered.

The composite diurnal cycle of moisture convergence and thermodynamic parameters is analyzed using the ECMWF operational model for the events with strong and weak convection during OTREC. Before the peak of convection at 12 UTC saturation fraction increases and instability index and DCIN decrease. Maxima and minima of saturation fraction, instability index and DCIN are reached simultaneously. This further confirms the importance of thermodynamic parameters that not only correlate in OTREC and PREDICT scatter plots, but show in the time evolution that they are responsible for convection development.

It is important to note that we are looking at convection and the environment rather than the evolution of individual plumes. Such analysis gives broad statistical answers and could help improve cumulus parametrizations, Raymond and Fuchs-Stone (2020).

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