# Abstracting ENSO Spatial Patterns' Impact on Atlantic Tropical Cyclone Seasonal Frequency

2012-05-14

#### Abstract

#### 1 Introduction

Understanding and predicting tropical cyclone (TC) activity is of significant scientific and societal interest. Pacific Ocean sea surface temperatures (SSTs) have well documented global long-range teleconnections, including Atlantic TC activity [9, 4, 5, 6, 12]. The quasi-periodic cycle (2-7 years) of warming and cooling of the near equatorial Pacific Ocean, known as the El-Niño Souther Oscillation (ENSO), is characterized by the warm El-Niño (EN) phase and a cold La Niña (LN) phase. Researchers have traditionally used the phase of ENSO to predict Atlantic TC activity, however due to the large amplitude variations in seasonal TC counts, the difference in Atlantic TC activity based on the various phases of ENSO is not obvious (see Figure 1).

ENSO has been quantified using warming-based indices where SST anomalies are averaged over regions in the Pacific. Such indices include the Nino 1+2 (0-10S, 90-80W), Nino 3 (5N-5S, 150-90W), Nino 4 (5N-5S, 160E-150W), and Nino 3.4 (5N-5S, 170-120W). Some studies have suggested such indices do not capture ENSO's nature and evolution. Subsequently, more elaborate indices were developed some of which were linear combinations of the above-mentioned indices [16], while others have proposed indices using transformed data or nonlinear combination of indices [13]. Despite the varying degrees of complexity, the majority of works attempting to capture ENSO focus on the intensity of warming in a given geographical region. While such indices might provide valuable insight into weather teleconnections (are there any good examples where ENSO does really well at predicting teleconnections?), they were not designed to capture the physical pathways by which Pacific SSTs may impact the large-scale conditions over the Atlantic.

In this paper we propose a novel spatial ENSO index that is designed specifically to capture the physical pathways by which Pacific SSTs may influence Atlantic TC activity. Our approach introduces a distance-based ENSO index that

tracks the location of the location of maximum near-tropical Pacific warming anomaly instead of its absolute warming. We will demonstrate the performance of our index by comparing it to traditional ENSO indices in both predicting seasonal TC frequency as well as discriminating between the large-scale conditions that are favorable for Atlantic cyclogenesis.

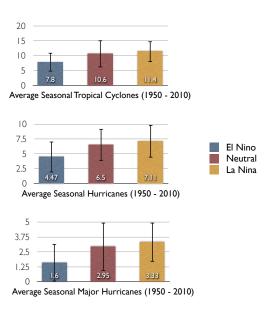


Figure 1: The 1950-2010 seasonal mean Atlantic tropical cyclones (top), hurricanes (middle), and major hurricane (bottom) counts for El-Niño, neutral, and La Niña years. Vertical bars denote standard deviation. The overlap between between bars across categories make distinguishing between Atlantic TC activity based on the phase of ENSO ambiguous.

#### 1.1 ENSO Overview

The quasi-periodic cycle (2-7 years) of warming and cooling of the near equatorial Pacific Ocean, known as the El-Niño Souther Oscillation (ENSO) is associated with anomalous atmospheric circulation and alterations to the Eastern Pacific thermocline (the subsurface boundary between upper warm waters and deep cool waters). During its warm, El-Niño (EN) phase, the equatorial Pacific Ocean experiences weak easterly winds causing an increase in Eastern Pacific SSTs, that in turn alters the atmospheric zonal (Walker) circulation, generally resulting in prevailing westerlies. ENSO's cold, La Niña (LN) phase, is characterized by the opposite atmospheric conditions – with cold SST anomalies along the Eastern Pacific and warm ones near the Western Pacific as a result of prevailing easterly winds (see Figure 2). The mechanisms that control the reversal to the opposite LN phase are not fully understood [11, 14]. Recent research

has suggested that to fully capture ENSO activity, it is no longer sufficient to monitor the warm and cold phases in the Eastern Pacific. Instead, warming patterns in the Central Pacific must be monitored as well [1]. Warming in the Central Pacific, known as El Niño Modoki, where a warm waters are surrounded by cold ones has been observed with increased frequency since the 1990s. Such changes have been attributed to anthropogenic global warming [19] as well as natural climate variability [18].

Enhanced convection as a result of anomalous Pacific Ocean warming is associated with strong westerly upper tropospheric wind over the Caribbean basin and tropical Atlantic, resulting in low TC activity during EN events and high TC activity LN [9]. Other studies have suggested that ENSO impact Atlantic TC activity via tropospheric warming [15].

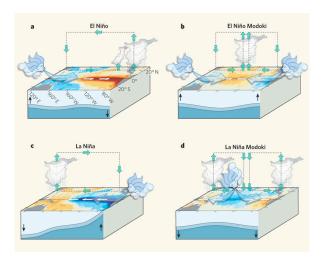


Figure 2: (a), An El Niño event is produced when the easterly winds weaken; sometimes, in the west, westerlies prevail. This condition is categorized by warmer than normal sea surface temperatures (SSTs) in the east of the ocean, and is associated with alterations in the thermocline and in the atmospheric circulation that make the east wetter and the west drier. (b), An El Niño Modoki event is an anomalous condition of a distinctly different kind. The warmest SSTs occur in the central Pacific, flanked by colder waters to the east and west, and are associated with distinct patterns of atmospheric convection. (c), (d), The opposite (La Niña) phases of the El Niño and El Niño Modoki respectively. Image and caption used for illustration purposes only taken from [2]

Acronym	Quantity
PWP-Lon	The longitude of the warmest SST pool in the northern near equatorial Pacific
PWP-Pres	The mean presure of PWP
PWP-OLR	The mean OLR of PWP
MinPres-Lon	The longitude of the region with the lowest mean central pressure
PWP-PCP	The longitudinal distance between PWP and PCP
Combo Index	A linear combination of xxxx

Table 1: A List of all quantities computed for this study with their corresponding acronyms.

# 2 Spatial ENSO Index (S-ENSO)

An increasing number of studies have suggested changes in the spatial warming patterns of the Pacific Ocean and some have linked those changes U.S. hurricane landfall probabilities [10]. We propose that based on such results, the spatial distribution of Pacific Ocean warming might provide better predictive insights into ENSO-Atlantic TC activity than warming anomalies alone. We propose a distance-based ENSO index that tracks the location of maximum near-tropical Pacific warming anomaly (Pacific Warm Pool or PWP thereafter) instead of its absolute warming.

For each season, we search for the maximum positive warming anomaly in regions of size comparable to that of warming-based ENSO indices. The PWP is selected by searching the Northern Tropical Pacific  $(0-30^{\circ}\ N)$ . The time series of the longitude of the PWP is then correlated with various quantities that communicate August-October Atlantic TC activity: number of tropical cyclones, number of major hurricanes, potential dissipation index (PDI) [7], accumulated cyclone energy (ACE) [3], and net tropical cyclone energy (NTC) [8].

In addition to tracking the longitude of the PWP, we also monitored other environmental factors that might influence Atlantic TC activity: the mean pressure value over the PWP, the mean Outgoing Longwave Radiation (OLR) of the PWP to monitor atmospheric deep convection, the longitude of the minimal pressure region to approximate Pacific cyclone activity, and finally the longitudinal distance between PWP and the coldest near equatorial Pacific SST region (Pacific Cold Pool or PCP). Finally, we ran a series of exhaustive experiments to linearly combine the above mentioned quantities into a single index (named Combo Index). Each index is built by computing the z-score for each quantity (i.e. mean OLR, etc.) then we add each normalized quantity to build a single index to use for analysis. Please see table ??

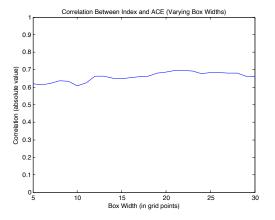
Tables ?? through ?? show the linear correlation coefficients between August-October TC Atlantic TC activity and our ENSO spatial index for Feb-April, April-June, and August-October respectively. In all cases our spatial ENSO index correlates better than traditional warming-based indices. The improvements increase as we increase lead time, with April lead times improving by

more than an order of magnitude. This improvement is because traditional ENSO indices suffer from a "predictability barrier" that make it difficult to use them to predict TC activity before June [17].

#### 2.1 Sensitivity Tests

To validate our results, we test the sensitivity of our results to the month ranges used to build the index, the size of the ENSO box used, the size of the search space, and the months used as TC season. As it can be seen from the figures below, our results are stable with respect to parameter choice, except for search space where the correlations chance drastically based on which region we search.

#### 2.2 Box Size Sensitivity Results



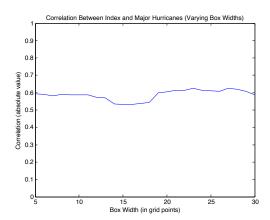


Figure 3: Corr Index vs. ACE

Figure 4: Corr Index vs. Major Hurricanes

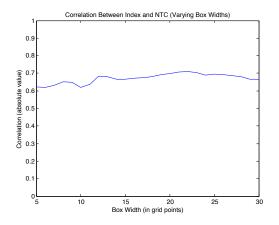
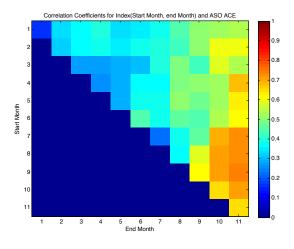


Figure 5: Corr Index vs. NTC

Figure 6: Corr Index vs. TCs

## 2.3 Varying Month Ranges



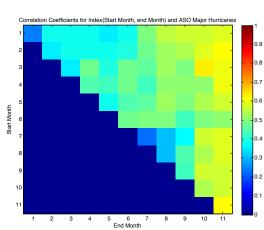


Figure 7: Corr Index vs. ACE

Figure 8: Corr Index vs. Major Hurricanes

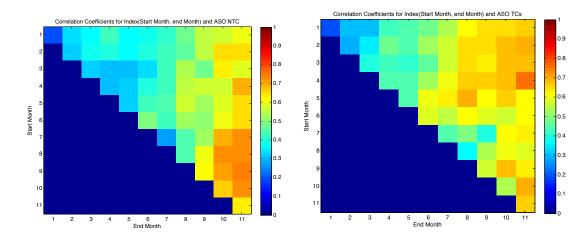


Figure 9: Corr Index vs. NTC

Figure 10: Corr Index vs. TCs

## 2.4 Varying Search Space (North and South Hemispheres)

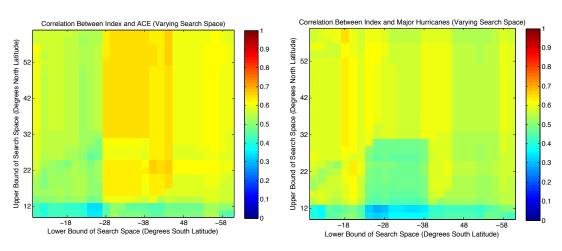


Figure 11: Corr Index vs. ACE

Figure 12: Corr Index vs. Major Hurricanes

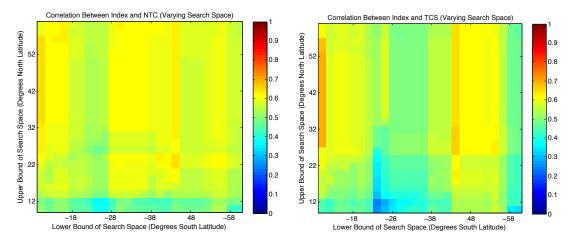


Figure 13: Corr Index vs. NTC

Figure 14: Corr Index vs. TCs

## 2.5 Varying Search Space (North Hemisphere)

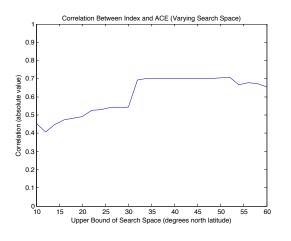


Figure 15: Corr Index vs. ACE

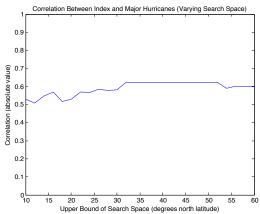
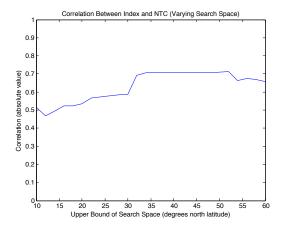


Figure 16: Corr Index vs. Major Hurricanes



Correlation Between Index and TCS (Varying Search Space)

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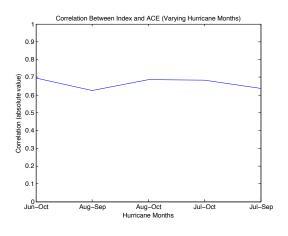
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Figure 17: Corr Index vs. NTC

Figure 18: Corr Index vs. TCs

# 2.6 Varying Hurricane Season



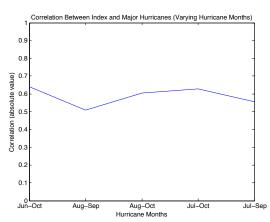
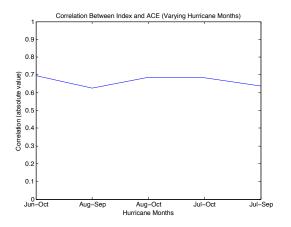


Figure 19: Corr Index vs. ACE

Figure 20: Corr Index vs. Major Hurricanes



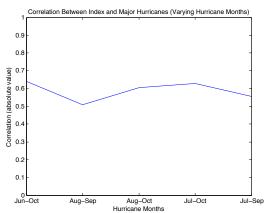


Figure 21: Corr Index vs. ACE

Figure 22: Corr Index vs. Major Hurricanes

# 3 Difference Composite Maps

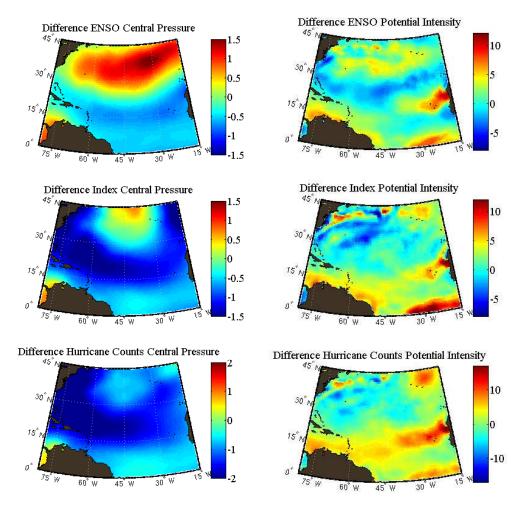


Figure 23: Diff Pressure Composites

Figure 24: Diff PI Composites

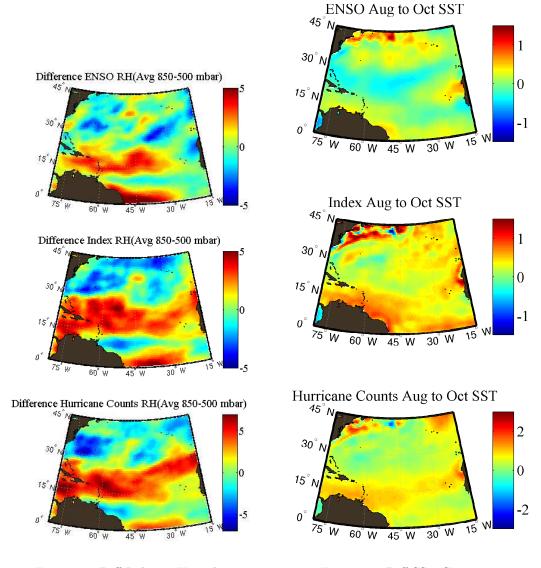


Figure 25: Diff Relative Humidity

Figure 26: Diff SST Composites  ${\cal S}$ 

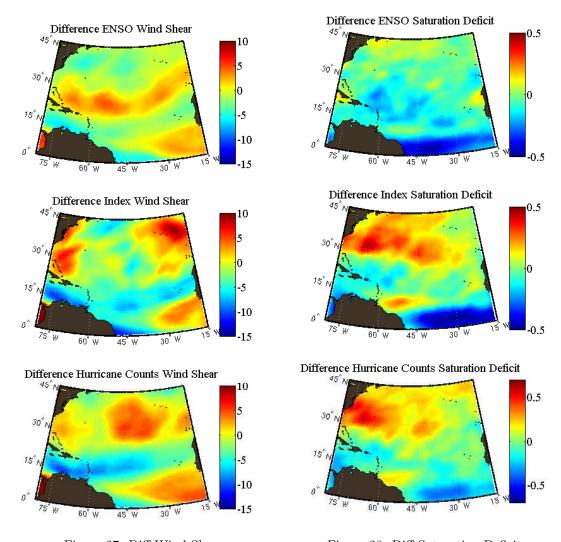


Figure 27: Diff Wind Shear

Figure 28: Diff Saturation Deficit

# 4 Average Difference Bar Graphs

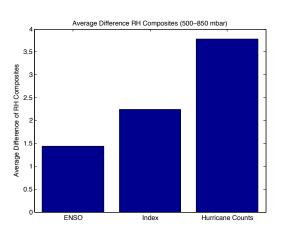


Figure 29: Avg Diff Relative Humidity

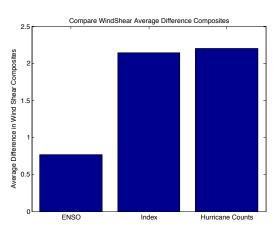


Figure 30: Avg Diff Wind Shear (between 850-200mbar)

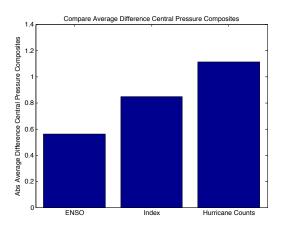


Figure 31: Avg Diff Central Pressure

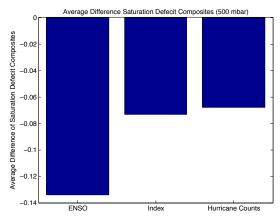
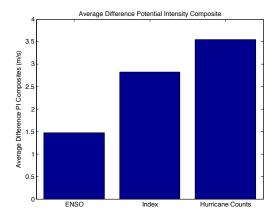


Figure 32: Avg Diff Saturation Deficit



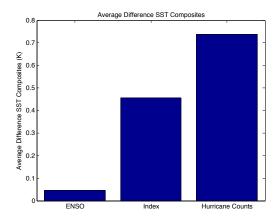


Figure 33: Avg Diff Potential Intensity

Figure 34: Avg Diff SST

## 5 Combination Index Composites

In this section, we compute four different indices, take the z-score of each, and then sum all of them up into one index. The four indices are: the OLR value inside the box of the warmest SST box, the pressure value inside of the warmest SST box, the longitude of the box with the lowest pressure, and the longitudinal distance between the warmest SST box and the coolest SST box. In this section we should the results where we get the years of the highest index, composite 5 different variables, do the same for the years of low index, and then subtract the composites.

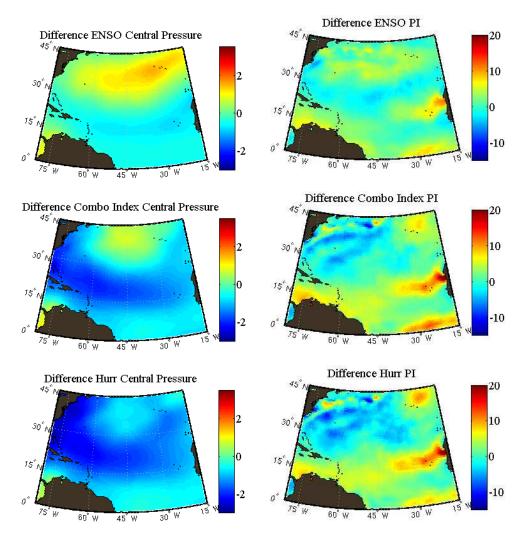


Figure 35: Diff Pressure Composites

Figure 36: Diff PI Composites

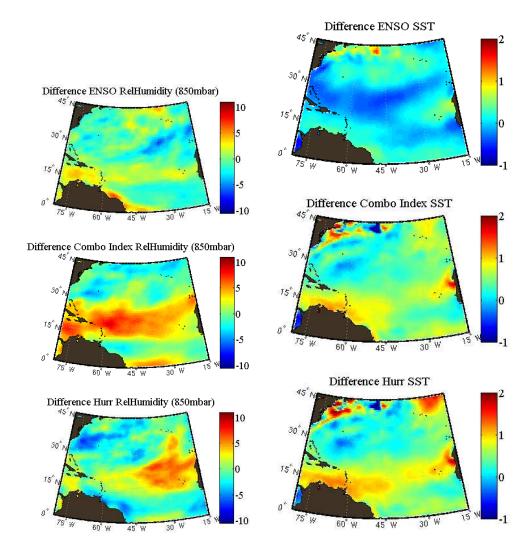


Figure 37: Diff Relative Humidity  $850~\mathrm{mbar}$ 

Figure 38: Diff SST Composites

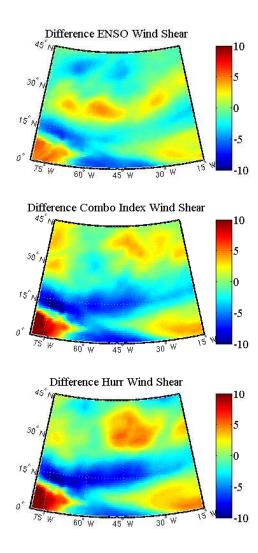


Figure 39: Diff Pressure Composites

## 6 Combo Index Cross Validation

In this section we compute the combination index in the same fashion as described int he previous section. These results are for building the index with varying month ranges, and then performing leave one-out cross validation with the combination index and 5 different hurricane statistics

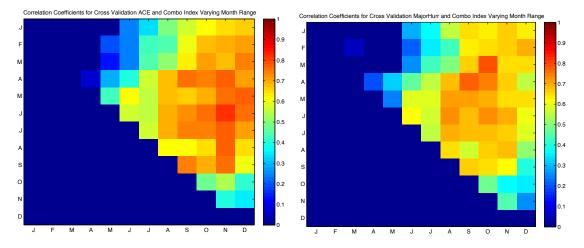


Figure 40: Corr Combo Index vs. ACE

Figure 41: Corr Combo Index vs. Major Hurr

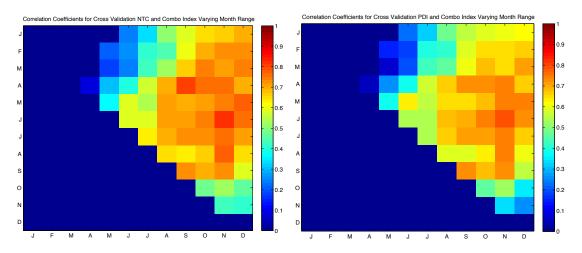


Figure 42: Corr Combo Index vs. NTC

Figure 43: Corr Combo Index vs. PDI

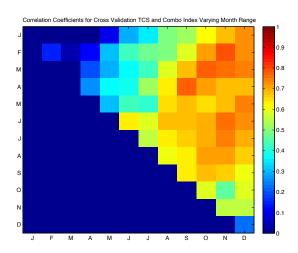


Figure 44: Corr Combo Index vs. TCs

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