This study examines relationships between available surface meteorology variables and climatic oscillations by using canonical correlation analysis (CCA). Canonical loadings and cross loadings from CCA are evaluated for meteorological stations located over coastal regions of Colombia. The tests, used for these studies, consider the temperature, the precipitation data, three of the main oscillations – the Ocean Niño Index (ONI), North Atlantic Oscillation (NAO), and the Quasi-biennial oscillation (QBO). The results show the power of the statistical method used to identify associations on the data set with an acceptable level of confidence using multivariate approach. The analysis reveals relations mostly between the variables and the ENSO for all cases and a discrete connection with the NAO and QBO. To add climate indices to the group of independent variables increased the variance rates between 5 and 7%.

Key words: Climate variability, atmospheric oscillations, Multivariate analysis, El Niño Southern Oscillation (ENSO)

I. INTRODUCTION

Colombia is located on the equatorial region in the northwestern of South America, bordering to the North with the Caribbean Sea; and to the west with Pacific Ocean. Colombia is the unique South American country which has both Caribbean and Pacific lowlands coastal regions. The Climate of Colombia is tropical and isothermal as a result of its proximity with the Equator, where the trade winds converge. On the other hand the Andes Mountains bring on convective processes and cause variety of climates according to altitude, other general aspects of the climate in Colombia can be found in [1], [2] and [3]. The Intertropical Convergence Zone (ITCZ) migration and El Niño South Oscillation are the main sources of seasonal and inter-annual climate variability in Colombia, however, the influence of other large scale climatic oscillations on the behavior of the surface meteorological variables is not yet fully understood, as mentioned in [4],[5] and [6].

The researches about connection between large scale climatic oscillations and tropical phenomenon contributes to a better understanding of the global climate and better seasonal-to-interannual forecasting, it is also affirmed by [4] and [7]. The main problem to advance in these studies lies in the limited availability of time series whose length allows to study interdecadal and interannual fluctuations. Despite the difficulty several authors have attempted to evaluate the influence of the macroclimatic oscillations in local weather patterns. Some examples are the study
of: the North Atlantic Oscillation (NAO) presented in [6] and [8], Quasi-biennial Oscillation (QBO) summarized in [1], [9] and [10]; and El Niño Southern Oscillation (ENSO) which has been the most studied, some representative papers about are [2], [3], [5] and [11].

The NAO index is used to represent this oscillation. It is defined as the normalized difference between the pressure at sea level in Gibraltar and southwest of Iceland Reykjavik. A positive index means an strengthening in the High Azores in the northern hemisphere causing an atypical change in the air temperature, increase in winter and decrease in summer. A negative index implies the opposite process, for more information about NAO consult [6], [8] or [12]. The cross correlation analysis between the NAO and the precipitation in Colombia, described in [6], [8], show a strong and negative association especially in the area of the Atlantic coast and a weaker and positive towards the center and south of the country with a six-month lag correlations; however, authors recommend pursue the subject.

The QBO index is used to represent the QBO phenomenon. It is the direction of zonal wind at 30 hPa level. The fluctuation in the wind direction is measured in the low and medium equatorial stratosphere with a period between 20 to 36 months and an average of 26 months. Some studies as [9], [10] and [13] indicate that QBO can affect the frequency of tropical cyclones in the Atlantic and therefore, the climatic conditions in the northwestern of South America (correlated the QBO phenomenon with the precipitation and the long-wave radiation from various areas of the country without obtaining statistical association, for that reason the authors suggest to use another type of correlation methods. Despite the existence of a quasi-biennial band in rainfall records in the country, there still are few published studies showing the possible link between QBO and hydrology of the country.

Finally, the Ocean Niño Index (ONI) represents the ENOS. The ONI is calculated as the running 3-month mean SST anomaly for the Niño 3.4 region, based on periods of 30 years updated every five. Values above 0.5 for three consecutive months are related to events El Niño and below -0.5 to Niña events. Studies as [2], [3], [5] and [11] show that the relation among ENSO, rainfall, and river flows is strongest during the December-January-February (DEF) and the weakest for the period March-April-May (MAM) These studies also reveal that the impact of ENSO on hydrological variables spreads from west to east. Unlike these studies, this paper proposes the use of canonical correlation between macroclimatic indices such as the ONI, NAO and QBO and variables as the maximum temperature and precipitation.

Canonical Correlation Analysis (CCA):

The CCA is a linear multivariate method used to compare two data sets, one independent \( X \) with \( p \) variables \( x_i \) and one dependent \( Y \) with \( q \) variables \( y_j \). The ACC creates linear combinations \( x^* \) and \( y^* \), called canonical variates (CV):

\[
x^* = X\alpha = \sum_{i=1}^{p} \alpha_i x_i \quad y^* = Y\beta = \sum_{j=1}^{q} \beta_j y_j
\]

The vectors \( \alpha \) and \( \beta \), known as canonical weights, must maximize the correlation between \( x^* \) and \( y^* \). Each CV must have non-zero variance and not correlated with another CV, that is:
\[ \text{Corr}[x^*_k, y^*_M] = \begin{cases} r, k = M \\ 0, k \neq M \end{cases} \quad (2) \]

Where \( r \) is the canonical correlation. The vectors \( \alpha \) and \( \beta \) and the \( r \) value are obtained by solving the coupled eigenproblem with the same eigenvalue \( \lambda \):

\[
[M_x] \alpha = \lambda \alpha \\
[M_y] \beta = \lambda \beta 
\]

Where \([M_x]\) and \([M_y]\) are the canonical correlation matrices:

\[
[M_x] = [S_{XX}]^{-1} [S_{XY}] [S_{YY}]^{-1} [S_{YX}] \\
[M_y] = [S_{YY}]^{-1} [S_{YX}] [S_{XX}]^{-1} [S_{XY}] 
\]

\([S_{XX}] \) and \([S_{YY}] \) are the variance and covariance matrix of the \( p \) elements of \( X \) and \( q \) elements of \( Y \) respectively. \([S_{XY}] \) and \([S_{YX}] \) are the covariance matrices between the elements of \( X \) and \( Y \). In this context \( \lambda \) represent the percentage of overlapping variance between the canonical variate pairs. Therefore \( \lambda \) is the canonical correlation squared and \( r = \sqrt{\lambda} \).

The canonical weights \( \alpha \) and \( \beta \) have the same interpretation as regression coefficients and are important when creating a predictive model. However the canonical loadings are considered useful to interpret the canonical relationships because they reveal how well do the variates are related to their own set of measured variables, it is justified in [14].

In addition to \( \alpha \) and \( \beta \), the method also compute the canonical loadings (CL) and canonical cross loadings (CCL) are also calculated. The independent CL are the linear correlations between \( x_i \) and \( x^* \), the dependent CL are the linear correlations between \( y_i \) and \( y^* \), the independent CCL are the linear correlations between \( x_i \) and \( y^* \) and the dependent CCL are the linear correlations between \( y_i \) and \( x^* \), the derivation of this concepts are in [14], [15]. The redundancy is also calculated to measure how much of the average proportion of variance of the original variables of one set may be predicted from the variables in the other set. The ACC was the method selected because it is the least dependent on the timing of the events, and in this case we are analyzing processes that involve different time scales, a similar analyses were carried out in [14].

II. METHODOLOGY

The CCA was implemented to relate the indices of the NAO, ENSO and QBO with the precipitation and maximum temperature obtained from two weather stations, one located in the Pacific coast, being in Buenaventura (Valle), and another one located in Cartagena (Bolivar) on the Caribbean Coast.
The data sets were obtained from the Instituto de Hidrología, Meteorología y Estudios Ambientales de Colombia. Weather stations were chosen for having multiyear monthly data over 30 years in both variables and for having the lowest number of missing data for the period 1982-2012. The indices were obtained from the website of the Earth System Research Laboratory (ESRL, 2014) for the period 1982 to 2012 also.

Time series of precipitation, maximum and minimum temperature were grouped into seasonal values as follows: December January, and February (DJF); March, April, and May (MAM); June, July, and August (JJA); and September, October, and November (SON). According to [15] divide series into seasonal values maximizes the probability to find relationships between climate index and meteorological variables. The dependent data set was composed of the seasonal series of precipitation, maximum and minimum temperature, and the independent of all four seasons of the climate indices.

To identify which combination of indexes shows the highest correlations two different tests using CCA were carry out. NAO and ONI index were the independent variables in first test and NAO, ONI and QBO in the second one. Both test were performed for each weather station and seasonal serie, it means that the CCA script was run 16 times.

F test was performed with the approach of Rao and statistically significant results were accepted with a confidence level of at least 90%. For each ACC were estimated the vectors $\alpha$ and $\beta$, 

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Fig. 1. Location of weather stations in the Pacific and Caribbean Coast. One being in Buenaventura (Valle-Colombia), and another one in Cartagena (Bolivar-Colombia)
canonical correlation (r), canonical loadings (CL) and canonical cross loadings (CCL), redundancy coefficients and reliability percentages. There is no a universal standard of significance to consider canonical correlations obtained from an ACC, so the expert must interpret the results regarding the nature of the physical processes and their own experience. According to [15] a threshold of 0.4 for dependent cross loadings, and 0.2 for independent cross loadings were considered. Besides any variable must share at least 16% of the variance with the respective variate and only the loadings that meet the 95% statistical significance are used.

The results were summarized in bar charts, the height of each bar is the product between the dependent loading multiplied and the independent cross loading that were statistically significant and above the thresholds mentioned. As it was mentioned above, the dependent loadings are the linear correlations between precipitation and maximum and minimum temperatures with its respective variate or canonical variable. And independent cross loadings are the linear correlation between the climate indexes and the dependent canonical variable (i.e. linear combinations of precipitation, maximum and minimum temperature). Multiplying these two loads allows identifies the degree of relationship between each index and the observed variables. For this work it was written a code in R version 2.14.2, which allowed to run all the 16 analyses. The code was written to evaluate the original data, to build the seasonal series, select the values according to the thresholds and display the results in bar charts.

III. RESULTS

The mean monthly precipitation, maximum and minimum temperature for both weather stations is showed in Fig. 1 and 2. Buenaventura is one of the rainiest places in the world with 6000–7000 mm per annum. Precipitation in Buenaventura has a bimodal annual cycle with two maximum around April-May and the second peak around October-November. This behavior corresponds with the displacement of the intertropical convergence zone (ITCZ). The mean temperature is 25.8 °C, with the lowest monthly average values during November to January and the highest during March to May. The average maximum and minimum temperature varies around 33.1°C and 19.7°C respectively.

The annual rainfall in Cartagena is around 912 mm per annum. Precipitation in Cartagena has only one long rainy season, from May through November. The mean temperature is 27.7 °C, with the lowest monthly average values during November to March and the highest during May to September. The average maximum and minimum temperature varies around 31.5°C and 24.2°C respectively.
Fig. 2. Mean monthly precipitation, maximum and minimum temperature from the weather station in Buenaventura (Valle-Colombia).

Fig. 3. Mean monthly precipitation, maximum and minimum temperature from the weather station in Cartagena (Bolivar-Colombia).

The canonical correlations and proportion of variance explained for test one are summarized in Table 1. The proportion of variance of meteorological variables explained by the NAO and ONI index rates between 24% and 42% in the period DEF, between 15% and 35% in MAM, between 22% and 30% in JJA and between 38% and 40% during SON.
Table 1. Canonical correlation coefficients (r) and variance explained for each CCA test.

<table>
<thead>
<tr>
<th></th>
<th>DEF</th>
<th>MAM</th>
<th>JJA</th>
<th>SON</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>r</td>
<td>v</td>
<td>r</td>
<td>v</td>
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<tr>
<td>Test one</td>
<td></td>
<td></td>
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</tr>
<tr>
<td>Buenaventura</td>
<td>0.86</td>
<td>0.42</td>
<td>0.66</td>
<td>0.35</td>
</tr>
<tr>
<td>Cartagena</td>
<td>0.7</td>
<td>0.24</td>
<td>0.57</td>
<td>0.15</td>
</tr>
<tr>
<td>Test two</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Buenaventura</td>
<td>0.9</td>
<td>0.55</td>
<td>0.7</td>
<td>0.46</td>
</tr>
<tr>
<td>Cartagena</td>
<td>0.75</td>
<td>0.36</td>
<td>0.8</td>
<td>0.30</td>
</tr>
</tbody>
</table>

The season with the highest percentages is DEF, followed by SON. The lower correlation coefficients and explained variance percentages were obtained in the case of Buenaventura and the largest in Cartagena. The difference in the results obtained in these two cases is explained by the contrast of the hydrological cycle between the two places. In Buenaventura the precipitation cycle is controlled primarily by the ITCZ but also for the equatorial west monsoon present in the northwestern of South America, as mentioned in [11]; In Cartagena rainfall response to the ITCZ migration and to the influence of the ocean–atmospheric processes in the Tropical Atlantic.

The numbers next to the label of the indices in Fig. 4 make reference to the period, 1 corresponds to DEF, 2 to MAM, 3 to JJA and 4 to SON. The follow results were common for all places: During DEF significant associations between meteorological variables with ONI1, ONI3 and NAO1y NAO3 were obtained. In season MAM only the variables from Cartagena showed to be related to NAO3. In JJA only maximum and minimum temperature from Cartagena showed to be related to NAO3. In season SON significant associations between meteorological variables with ONI1, ONI2 ONI3, ONI4 were obtained.

In general, the higher canonical loadings occur between the ONI index and the variables over DEF in Cartagena. And the lowest values are obtained between the NAO and variables of Buenaventura. Season MAM is the period with non-significant relations among large scale oscillations, temperature and precipitation in coastal regions of Colombia. The results are according with [2], [6] who obtained the highest significant cross-correlations between ENSO and hydrological variables during DEF season, and non-significant for period MAM.
The results also correspond with the physical phenomenon, for example, a positive ONI (associated with the Niño event) is related to a decrease in rainfall and a rise in maximum temperature. It also happens that the relationship between ONI and meteorological variables has a three to six-month lag. This corresponds to the fact that the ENOS spreads from west to east and it takes across the Pacific equatorial about six months.

The relation between NAO index and meteorological variables is coherent with results in [15]. They report that the influence of the NAO changes depending on time of year and do not follow a definite pattern. The relationship with NAO can be understood considering that the NAO affects the intensity of trade winds from the northeast and therefore the displacement of the ITCZ over the tropical Atlantic, modifying the moisture transport from the tropical Atlantic to the Americas and convective activity. The effect of NAO on-year band on precipitation appears to be limited.

In the second test, with QBO in the group of independent variables, the proportion of variance of meteorological variables explained by the indices was slightly higher. Values ranged between 36% and 55% in the period DEF, between 30% and 46% in MAM, between 35 and 40% in JJA and between 38% and 41% during SON.

The bart chart obtained for the second test was similar to Fig. 4, the climate oscillation with higher connection to the variables was the ONI. In all cases the relationship was directly and indirectly with the maximum temperature with the precipitation, with greater weight for DEF and SON. As for the NAO canonical charges had the same meaning as in the first test but were lower in magnitude. The only difference with the first test was the sign of the QBO appeared slightly related to precipitation.

This slightly additional relationship that appears between QBO and seasonal series explains the slight increase between 3 and 7% of the variance explained by climatic oscillations. The QBO

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**Fig. 4. A graphical depiction of the CCA results for Test One**
reaches its greatest extent in the stratosphere, the most accepted hypothesis is that its existence is
associated with the displacement of equatorial Rossby-Gravity and kelvin waves. The QBO is
related to the frequency of cyclones in the Atlantic or to the ENSO cycle, so it is possible to observe
the signal in Colombia. The importance of this work is that the results found by CCA are consistent
with the physical processes and correlation with previous studies performed. This implies that the
method is an alternative to identify simultaneous relations as also suggest [14]. The canonical
correlations are not very large because the variance explained by climatic oscillations is generally
small.

IV. CONCLUSIONS

The work shows that the CCA identifies partnerships through a multivariate approach of
macroclimatic indices as predictors. The consistency of the results found with the physical processes
and preliminary studies of correlation. In general the results of both tests reaffirm the influence of
ENOS on both Caribbean and Pacific lowlands coastal regions, but lower in the Pacific region. The
relationship is inverse to precipitation and direct with the maximum temperature; the highest
correlations occur over DEF and the weakest during MAM. The relationship between the climate
index and meteorological series is provided by the ONI, followed by the NAO and a more discreet
contribution of the QBO was obtained.

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