

The relationship between ENSO and Paraná River flow

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Abstract. Several studies indicate that there is a relationship between the climatic variability in the South American continent and alterations of the position and intensity of the heat sources in the equatorial region. The El Niño phenomenon can influence the precipitation over some regions of South America such as the Brazilian Northeast, Amazonia, South of Brazil and Uruguay. Over 80% of Brazil's energy comes from hydropower, and decisions concerning future availability and pricing require forecasts of river flow, ideally several months in advance. In this work the relationship between the Paraná River flow and the ENSO (El Niño/Southern Oscillation) mode is investigated and statistical forecasts of river flow are tested. An evaluation of the relationship between the Pacific sea surface temperature and the Paraná River flow indicates an ENSO pattern over the equatorial Pacific. The time series of the ENSO mode obtained by applying principal components analysis on the sea surface temperature (SST) were used as predictors for the Paraná River flow forecast. Improvement in the model forecast skill is also obtained by considering the lagged river flow time series as a predictor.

1 Introduction

The El Niño phenomenon influences the precipitation over parts of South America, such as Brazilian Northeast, eastern Amazonia, South of Brazil, Uruguay and northeastern Argentina. Sea surface temperature (SST) positive anomalies in the equatorial Pacific favour precipitation increase in the South of Brazil (Grimm et al., 1998; Coelho et al., 2002) and precipitation decrease in the Brazilian Northeast (Rao and Hada, 1990; Moura and Shukla, 1981). Robertson and Mechoso (1998) detected a significant relationship between a quasi-decadal component in the Uruguay and Paraná rivers flows and variations in the Atlantic SST, beyond of the interannual maximums in the El Niño – Southern Os-

cillation (ENSO) time scale. Robertson et al. (2001) analyzed the interannual to decadal predictability of the Paraná River extracting near-cycling components of the summer river streamflow. They found that the ENSO oscillatory component was associated with changes in the probability distribution of monthly flows and that the decadal modulation of ENSO may be important, although the predictability due to ENSO at interannual lead times is small. The averaged flows observed in the Paraná River during El Niño events are always larger than those observed during La Niña events (Berri et al., 2002).

The Paraná River is about 4900 km in length, being the second longest in South America draining all the centre-south of South America. More than 80% of Brazil's energy is provided from hydropower generation. Decisions relating to energy price and investment planning can be made with greater confidence if reliable forecasts of future volumes of water available for power generation can be calculated. This requires forecasts of future inflows into hydropower reservoirs. Several studies showed that the SST anomalies in the tropical Pacific can be used to forecast river flow in the North and Northeast regions of Brazil (Hastenrath, 1990) and Argentina (Berri and Flamenco, 1999).

The purpose of the present paper is to investigate the relationships between the Paraná River flow and the ENSO mode and to develop a statistic model to forecast Paraná River flow by using the time series of the ENSO mode.

2 Data and methodology

In the work reported here, the data set basis comprise 52 years (1950–2001) of natural flows into the Posadas reservoir from the Paraná river (27.36° S, 55.90° W) and the corresponding monthly SST for the Atlantic and Pacific Oceans (denoted by AO and PO, respectively) with spatial resolution of 1° × 1° (Reynolds e Smith, 1994). Using the 1° resolution SST data, mean SST at 3° intervals were calculated over ocean blocks. The latitudinal domain is restricted between

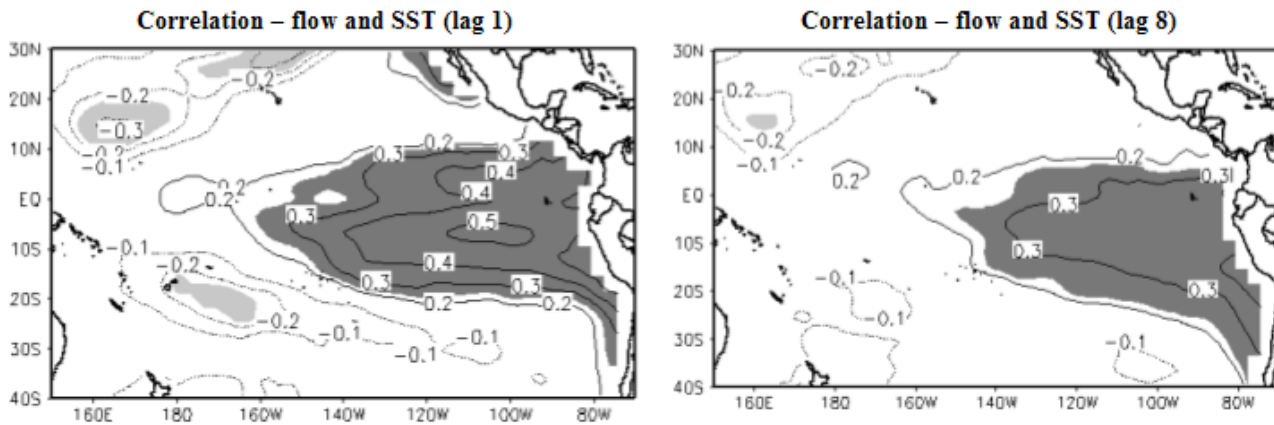


Fig. 1. Isolines of correlations coefficient between monthly natural Paraná River flow and SST lagged of 1 and 8 month. Shading represents significant correlation. Degrees of freedom (df) for each correlation is 50. Tabulated values for significant correlation at 5% level (t test) are 0.231.

30° N to 40° S for both AO and PO; for the AO the longitude range is 60° W to 20° E, and for the PO, 150° E to 70° W. The annual cycle was removed by subtracting from each observation (whether monthly flow or SST) the corresponding monthly mean and then scaling this difference by the monthly standard deviation. The “Pearson’s r ” correlation coefficient was determined between the time series of the Paraná River flow and SST for the PO lagged from one to twelve months.

The principal components analysis (PCA) reduces a data set containing a large number of variables to a data set containing fewer new variables, but representing a large fraction of the variability contained in original data (Weare and Nasstrom, 1982). The goal of applying PCA to a data set representing spatial and temporal variations is to decompose the original data set in modes represented by spatial patterns, time variation and the explained variance for each mode. PCA was used to reduce the dimensionality of the SST matrices and to find the prevailing modes of the ENSO. The physical interpretation of principal components (PC) is limited by the imposed spatial orthogonality of PC patterns (loadings) and the resulting temporal independence of the time coefficients (scores) (Richman, 1986). A rotation of the PC may help overcome this difficulty. The Varimax rotation was performed to find a mode characteristic of the ENSO pattern and to compare this PC mode with rotated principal component (RPC) mode, both associated with ENSO.

Canonical correlation analysis (CCA) is a statistical technique that identifies a sequence of pairs of patterns in two data sets (Wilks, 1995). CCA was applied to the PO and AO SST fields to identify the principal canonical mode that shows the coupling of the ENSO pattern with the AO SST. Each canonical mode is expressed with two spatial patterns (canonical maps for each field), two times series (scores for each field) and the correlation between the two times series.

A linear regression model was developed for forecasting the Paraná River flow from PC, RPC and canonical mode (CM) scores that are associated to modes with ENSO spa-

tial pattern in the tropical PO. Given the real time goal of the forecasting technique (i.e., flow forecast at time t with lag k), the statistical model takes into account data collected at previous times ($t-k$). However, any forecast model that uses SST data as input may be updating monthly, give that additional SST data as well as monthly flow data become available. This way, updating forecasts of future river flow requires that the SST scores be updated every month, even if the model parameters and modal structure are not. In this work, the first 42 years (504 months) of the full 52-year period were used to provide initial estimates that were subsequently updated month by month. The model skill was evaluated for 10 years of forecasts and the best predictors were identified by the stepwise regression. The parameters calculated to evaluate the model skill are: i) correlation coefficient (r) between the forecast and observed series; ii) square root of the mean squared error (RMSE); iii) mean error (Bias).

3 Results

Lagged correlation coefficients maps between river flow and PO SST’s show significant relationship up to lag 8 (months). The spatial pattern of correlation is similar to the ENSO pattern in the tropical PO (Fig. 1), confirming the ENSO influence in the Paraná river flow and its potential use as a predictor. Thus, both PCA and CCA were used to find the time variations of the ENSO pattern.

The first and second non-rotated SST PC show a spatial pattern similar to the ENSO pattern in the tropical PO (Fig. 2a). The first PC, which accounts for 34% of the total variance of PO SST, shows time fluctuations of low frequency with maximum (minimum) in El Niño (La Niña) years (Fig. 2b) and the Niño 3 area is shown to be more important. The second PC explains 11% of the total variance of the data set and presents variations at interannual and interdecadal scales (Fig. 2b). The second PC may represent an ENSO phase, similar to the Niño 1 and 2 areas from the point

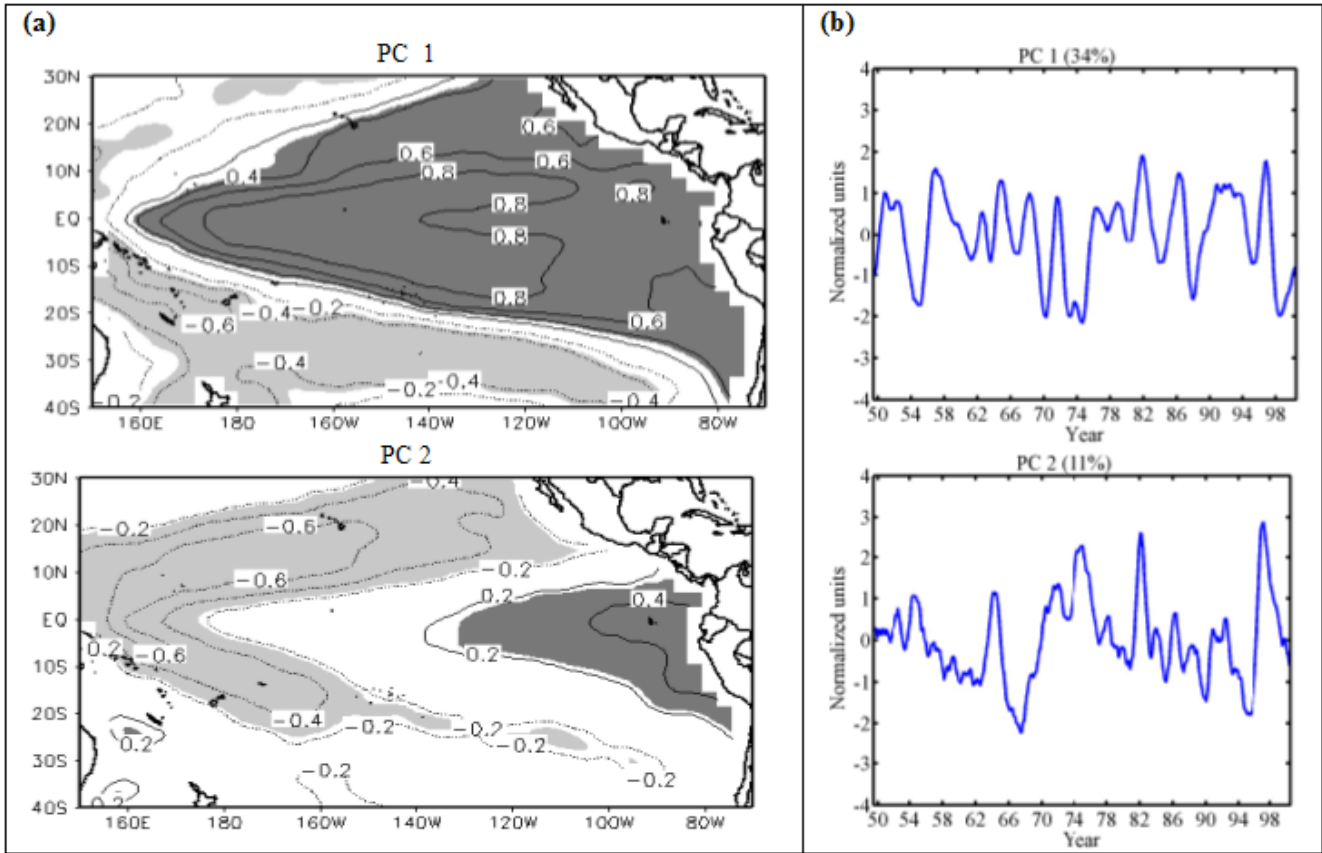


Fig. 2. (a) Spatial patterns of the first two PC modes of Pacific SST, presented as map correlation. (b) Times series of the first two PC scores. The times series are smoothed using 13-month running mean and the amplitude is normalized by the standard deviation.

of view of the SST anomalies. This pattern may later tend to the one represented by the first PC which has the largest SST anomalies in the Niño 3 area.

Figure 3a shows the first RPC of SST in the PO, showing a more localized spatial pattern, while the first PC covers almost all the tropical PO. The time series of the the first PC and the RPC are similar (Fig. 3b). However, after PC rotation, just one pattern remained statistically significant, with clear ENSO characteristics.

The CCA yields one pair of canonical pattern that shows the coupling of the ENSO pattern with the AO SST. This coupling is presented in the first CM and indicates a positive correlation between SST anomalies in the tropical PO and north tropical AO (Fig. 4a). In the south AO there is a dipole pattern with significant correlation with the ENSO area. The correlation between the two times series associated to the canonical patterns is 0.80 (Fig. 4b).

Linear regressions were calculated between the Paraná standardized anomaly of the river flow and the times series (scores) from the two first PC, from first RPC and from the first canonical mode. All four scores are associated to ENSO patterns and lagged by up to four months. Thus, the flow forecast was performed up to four lags. The goal is to find the best predictor and to evaluate the relative role of the ENSO pattern as a predictor of the Paraná river flow.

Table 1. Ordering of best predictors and the values of SSE to each lag of forecast flow obtained from the first and last forecast ($t=505$ and $t=624$, respectively).

Sum of Square Error (SSE) – $t=505$				
Predictor	Lag 1	Lag 2	Lag 3	Lag 4
PC 1	450.05	447.60	453.73	457.41
RPC 1	461.15	460.21	464.92	466.95
PC 2	484.64	484.91	484.05	484.24
CM 1	485.33	485.25	485.30	485.47
Sum of Square Error (SSE) – $t=624$				
Predictor	Lag 1	Lag 2	Lag 3	Lag 4
PC 1	547.12	543.12	549.20	554.27
RPC 1	563.53	561.77	566.93	570.14
PC 2	598.70	597.62	600.22	601.29
CM 1	603.20	603.24	603.23	603.20

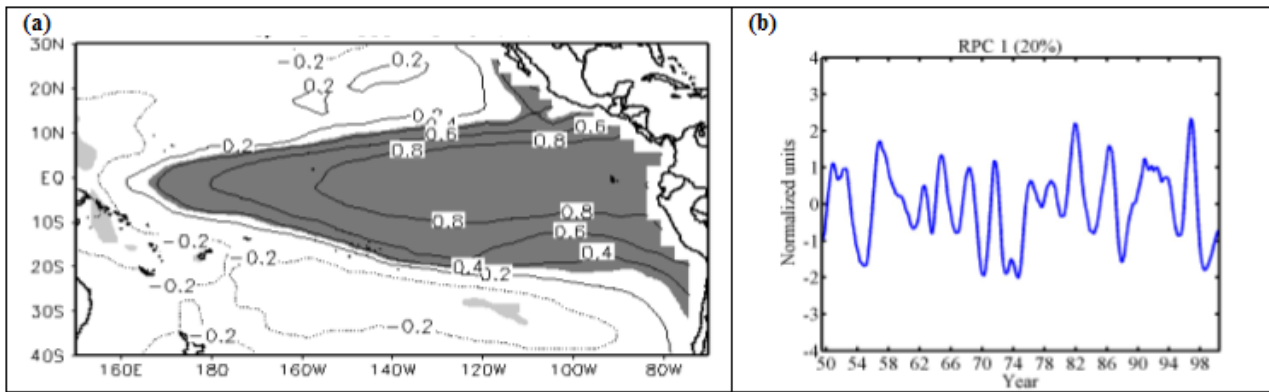


Fig. 3. (a) Spatial pattern of the first RPC modes of Pacific SST, presented as map correlation. (b) Time series of RPC scores of the first mode of the Pacific SST. The time series is smoothed using 13-month running mean, and the amplitude is normalized by the standard deviation.

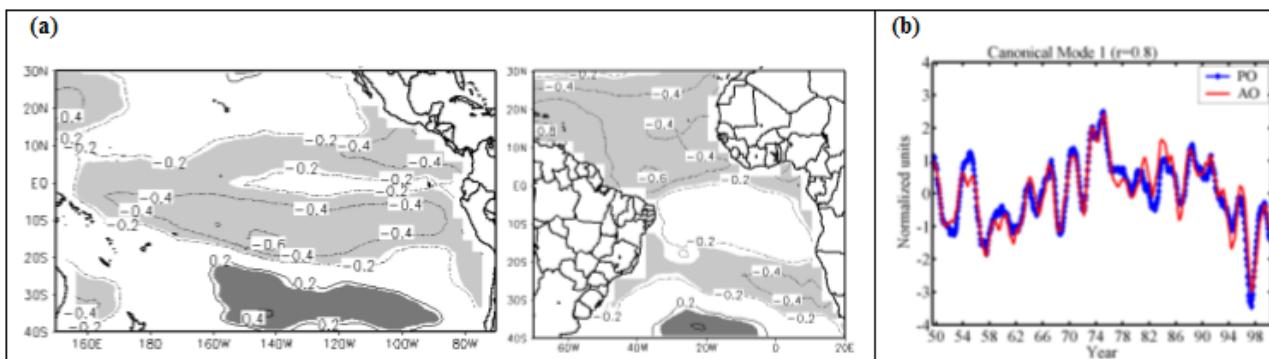


Fig. 4. (a) Spatial patterns of the first canonical mode of Pacific and Atlantic SST, presented as map correlation. (b) Times series of scores of the first canonical mode of the Pacific and Atlantic SST. The time series is smoothed using 13-month running mean and the amplitude is normalized by the standard deviation.

The stepwise regression indicated the score of the first PC as the best predictor to represent the ENSO mode (Fig. 2), up to lag 4. Table 1 shows the order of the better predictors and the sum of square error (SSE) to each lag of forecast flow obtained to the first and last forecast time ($t=505$ and $t=624$, respectively). The best forecast is obtained when the PCA identifies the ENSO pattern. There is a small difference between the SSE of a forecast obtained using the first PC and the first RPC, but this difference is larger when compared with the CM. This result suggests that the ENSO influence on the Paraná flow is direct and not indirect, through the AO, at least up to four months lag. Another important characteristic of the statistical forecast model is that the SSE is lower at lag 2 indicating thus that the best forecasts can be obtained two months in advance.

Figure 5 shows forecasts obtained with the regression model defined above, for the verification period (1992 to 2001). This figure shows the observed natural Paraná inflows (red solid line) together with their predicted values, from one up to four months lag (colored dashed lines). Only the first PC as a predictor is chosen by the stepwise regression. The skill parameters for each lagged forecast are presented in the

top of each map. In agreement with the stepwise regression, the model skill increases to lag two and decreases with increasing lead-time of the forecast. The annual cycle and the low frequency variations are well captured by the ENSO mode. Correlations between the forecast and observed series are significant, indicating that a considerable ratio of flow variance is explained by the ENSO. Figure 5 shows that the positive extreme values of river flow were severely underestimated at all forecast lags. In some cases the predicted value is less than half the observed value, and the high frequency variations are not well reproduced. However, the low frequency tendency is well captured.

4 Conclusions

It has been shown that multivariate analysis, such as PCA and CCA, are useful tools to find spatial and temporal patterns that describe the variability of SST anomalies. The ENSO patterns are captured in the first PC and first RPC. The coupling mode between the ENSO pattern and the Atlantic Ocean SST anomalies are also captured in the first mode.

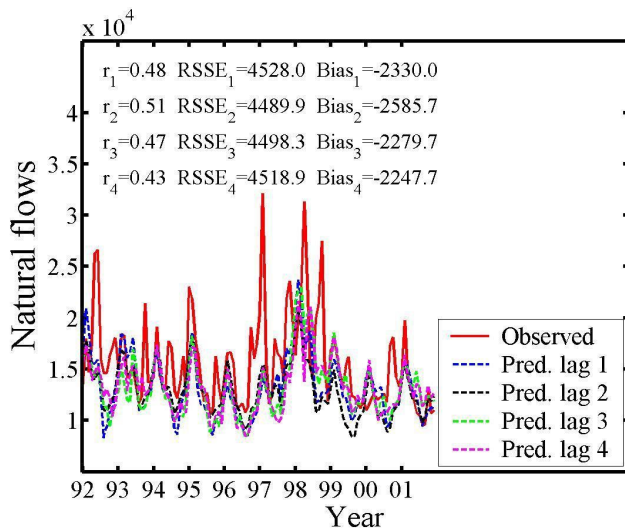


Fig. 5. Observed and predicted natural Paraná flow, updated monthly, from the scores of the first Principal Components of PO SST. Predictions are made with lead-times of from one to four months.

The best form to represent the ENSO mode and to use it as a predictor of river flow is by the first PC of the Pacific Ocean SST anomalies, in comparison with RPC and CCA. The better skill of the regression model was reached at lag 2 (i.e. two months in advance). The model skill, based on the evaluation metric (correlation coefficients, square root of the mean squared error and bias) is satisfactory, although extreme positive values are not well predicted.

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