Clustering of Loop Current patterns based on the satellite-observed sea surface height and self-organizing map

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The self-organizing map is used to investigate variations of the Loop Current (LC) in the Gulf of Mexico from 1992 to 2013 based on satellite-observed sea surface height data. It is found that LC variations can be characterized by three spatial patterns: normal, extension and retraction. The corresponding temporal variations confirm that LC eddy shedding generally occurs during the transition from the extension to retraction patterns. On the weekly time scale, the wind stress curl (WSC) in the Caribbean Sea has a major influence on LC eddy shedding. The increase of Caribbean WSC from June to November favours more frequent LC eddy shedding during that period. On the interannual time scale, there is also a potential linkage between the frequency of LC eddy shedding and El Niño activities.

1. Introduction

The Loop Current (LC) is the dominant ocean circulation system in the Gulf of Mexico (GoM) (e.g., Oey, Ezer, and Lee 2005). It originates at the Yucatan Channel and exits through the Florida Straits (Figure 1). One of its most notable characteristics is that it episodically sheds large, warm-core eddies which affect various aspects of GoM hydrodynamics. LC eddy shedding consists of highly nonlinear extension and retraction processes (e.g., Oey, Ezer, and Lee 2005). Sometimes, the LC can shed eddies without extension, and detached eddies can also re-attach to the LC. The time interval between eddy shedding events is found to be irregular, ranging from 0.5 to 18 months (Leben 2005).

The causes of the complex LC shedding process are still subject to debate. Hurlburt and Thompson (1980) pointed out that LC eddy shedding was controlled by the horizontal instability of the LC. Maul and Vukovich (1993) found that the ensemble correlation between monthly position of the LC and volume transport is zero based on 12 years of satellite and in situ data. Pichevin and Nof (1997) concluded that eddy shedding is required to satisfy the momentum balance principle. Oey, Lee, and Schmitz (2003) suggested that Caribbean Sea eddies spun up by local winds could propagate through the Yucatan Channel into the GoM and influence the LC eddy shedding process. Lugo-Fernández (2007) examined the LC through a dynamical system approach and found that the LC is nonlinear, but not chaotic. Nürnberg et al. (2008) explored the variability of the LC and its relation with Mississippi River discharge through a geochemistry view. Lugo-Fernández and Leben (2010) provided the linear relationship between LC retreat latitude and eddy separation period using the satellite altimeter-derived LC metrics. Sturges, Hoffmann, and Leben (2010) further proposed that the downstream Florida Strait...
Transport variation may trigger LC eddy shedding, while Oey and Chang (2011) argued that several model solutions showed that such a downstream trigger was not necessary. Numerous other numerical modelling efforts (e.g., Lee and Mellor 2003; Yin and Oey 2007; Chang and Oey 2012; Le Hénaff et al. 2012; Chang and Oey 2013a; Xu et al. 2013) were also made to depict the evolution of the LC. A recent study by Liu et al. (2012) suggested that the local net heat flux and the intensity of the advective heat flux convergence can introduce instability of the LC and associated eddies, highlighting that the complex coupling between atmosphere and ocean can contribute to the LC shedding (e.g., Chang and Oey 2013b).

In this study, we used a novel feature extraction method to characterize satellite-observed SSH data of the LC and further analyse its variations. The frequency of occurrence (FO) for the LC retraction pattern was then correlated with WSCs and several climate indices at different time scales.

2. Data and methods

Twenty one years (1992–2013) of gridded satellite altimeter-observed SSH data around the LC area (red box in Figure 1) were analysed. Altimeter data, distributed by the Archiving, Validation and Interpretation of Satellite Oceanographic (AVISO) data service, have spatial and temporal resolutions of 1/3° and 7 days, respectively. For quality control, only the data located inside the areas with water depth greater than 100 m were chosen (Liu, Weisberg, and Yuan 2008; Yin et al. 2014). In addition, wind data (with 3 hourly time interval and ~33 km spatial resolution) were extracted from National Centers for Environmental Prediction North American Regional Analysis (NARR). Weekly mean wind stress was subsequently calculated based on Cushman-Roisin and Beckers’s (2011) method, and then used to derive corresponding WSCs. Climate indices were
obtained from the Physical Sciences Division of Earth System Research Laboratory, National Oceanic and Atmospheric Administration (NOAA).

Our focus here is to apply an advanced method, the self-organizing map (SOM), to re-examine the long-term satellite observations archived for the LC region in the last two decades. Actually, satellite data have been widely used in studying the Gulf circulation and LC dynamics (e.g., Leben 2005; Leben and Honaker 2006; Lugo-Fernández and Leben 2010). Based on an unsupervised artificial neural network, the SOM is an effective method for feature extraction and classification, and can map high-dimensional input data onto the elements of a regular, low-dimensional array (Kohonen 2001). It has been demonstrated to be more powerful than the conventional empirical orthogonal function method for feature extractions, especially when the signal is highly nonlinear (Reusch, Alley, and Hewitson 2005; Liu, Weisberg, and Mooers 2006). The SOM has been shown to be a valuable tool in oceanographic studies (Liu and Weisberg 2011). It has been applied to identify patterns in ocean currents and sea surface temperature fields on the West Florida Shelf (Liu and Weisberg 2005; Liu, Weisberg, and He 2006; Liu, Weisberg, and Shay 2007), biogeochemical properties in the northern Adriatic Sea (Solidoro et al. 2007) and current variability in the China Seas (Liu, Weisberg, and Yuan 2008; Jin et al. 2010; Tsui and Wu 2012; Yin et al. 2014).

In this study, all weekly SSH data within the study domain were fed into the SOM as inputs. Based on the minimum Euclidean distance and pattern size given initially, different SSH patterns were extracted in a topology-preserving way. Each weekly SSH snapshot was then assigned to one of these patterns (e.g., Liu and Weisberg 2005). From these assignments, the best matching unit (BMU) time series and FO of each pattern were obtained. The SOM parameters such as lattice, weights, training method and neighborhood function were chosen according to Liu, Weisberg, and Mooers (2006). Based on our sensitivity experiments and the variability of the LC, the pattern number 1 × 3 with contrasting difference between each pattern was chosen prior to the training process.

3. Results

3.1. Spatial variability

The three SOM patterns and their corresponding FOs are shown in Figure 2. The 0.45 m SSH contour line was chosen as the edge of the LC and its detached eddies in this study, based on the examination of patterns and evolutions of LC and LC eddies over 21 years SSH data record. It is consistent with methods used by earlier studies (e.g., Leben 2005) in SSH contour selection. As shown in Figure 2(a), Pattern 1 (P1) is the LC’s normal condition, without extension or shedding. The north and west edges of LC in P1 reach about 26.5°N and 88°W, respectively. For P1, the LC is featured with significantly high sea level, accompanied by low sea level on the north-west edge. Pattern 2 (P2) is the LC extension pattern. The most obvious feature for P2 is that the LC extends into a relatively elongated shape, such that its north and west edges reach about 27.5°N and 90°W. In contrast, Pattern 3 (P3) represents the LC retraction pattern, which can also be considered as the eddy shed pattern. The main body of the LC and LC eddy are clearly separated from each other. After eddy shedding, the north and west edges of the LC retreat to about 25.5°N and 86°W. The shed eddy is located at about 26°N, 90°W with lower SSH and weaker geostrophic velocity than the main LC. Different from P1 and P2, P3 shows the LC further to the south-east after eddy shedding. There is also a large cyclonic eddy with low SSH present between the LC and its shed eddy.
3.2. Temporal evolution

Figure 2(b) shows the temporal changes (BMU time series) of the three LC patterns (P1, P2 and P3). A repeated cycle is generally evident in the 21-years’ time series. Using P1 as an arbitrary beginning, the cycle of P1→P2→P3→P1 is fairly robust. Specifically, the LC starts in its normal pattern (P1), and then extends to the north-west to 27.5°N to reach its extension pattern (P2). An eddy shedding event subsequently occurs, and then the LC...
retreats to about 25.5°N to its retraction pattern (P3). However, due to the nonlinearity associated with LC flow and vorticity dynamics (e.g., Lugo-Fernández 2007), not every warm eddy shedding process follows this cycle. In a few events (e.g., in 2005), there was no eddy separation followed by a proceeding extension pattern. In several other events (e.g., in 2008), the eddy separation pattern occurred right after the normal pattern, bypassing the extension pattern.

In order to quantify the percentage occurrence of each pattern, the FO was calculated by summing the number of occurrences of that pattern divided by the total record length. Over the 21-year study period, 41.5% of the LC patterns belong to the normal pattern (P1), while the extension (P2) and retraction (P3) patterns account for 28.5% and 30.0%, respectively. To better illustrate the seasonal variation of the LC and the dominant pattern for each month, the monthly FOs (MFO) of the three patterns were also calculated (Figure 2). The MFO of P1 is larger than those of P2 and P3 from January to May, which suggests that generally the LC tends to remain in its normal pattern (P1) during this period. In July and August, the MFO of P2 is the largest, suggesting that LC tends to extend during this period. The MFO of P3 becomes dominant from September to December, showing that retraction of the LC is more likely to occur during this period. We note that transitions from P1 to P2 and from P2 to P3 occur in June and the end of August, respectively, indicating that statistically the LC extension (eddy shedding) tends to occur in June (in late August).

4. Discussion

To further study the LC eddy shedding process and its possible mechanisms, we compared the weekly and annual FOs of P3 with long-term mean wind stress curl (WSC) over different spatial domains, as well as with different climate indices.

Figure 3(a) shows the weekly FO (WFO) of P3 along with long-term weekly mean WSC of three (Caribbean Sea, Bahamas and GoM; see Figure 1) previously identified wind influence regimes (e.g., Oey, Lee, and Schmitz 2003; Sturges, Hoffmann, and Leben 2010; Gopalakrishnan, Cornuelle, and Hoteit 2013). We found that the zero time lag correlation coefficient between CS WSCs and WFO of P3 is 0.83, much higher than the correlations with the local WSC in the GoM (correlation coefficient $r = 0.45$) and the downstream WSC in the Bahamas area ($r = 0.65$). This suggests that LC eddy shedding is likely associated more with CS WSC in the upstream Caribbean Sea. Indeed, Oey, Lee, and Schmitz (2003) showed that negative CS WSC can spin up Caribbean eddies (anticyclones), which in turn lead to a lower frequency of LC shedding. Our results further reveal the CS WSC increases from June to November. It can play a role in suppressing the formation of anti-cyclonic eddies in the CS. In other words, the increase of CS WSC during this period favours a higher frequency of LC shedding, as shown in Figure 3(a).

To understand the interannual variability of LC eddy shedding, the annual FO (AFO) of P3 was examined together with various climate indices, including the ONI, North Atlantic Oscillation index (NAO), Southern Oscillation Index (SOI) and Pacific Decadal Oscillation (PDO). Various time lag correlations and running averages were performed. Among them, the best correlation ($r = 0.6$ at 95% confidence interval) is found between the six-month moving-averaged ONI and the AFO of P3 with a 90-day lag (Figure 3(b)). Other indices show no significant correlations (not shown). The ONI is defined as the sea surface temperature anomalies in Niño Region 3.4 (5°N–5°S, 120°W–170°W) and used as an index for El Niño (e.g., Kousky and Higgins 2007). The relationship between ONI and AFO of P3 suggests a possible connection between Pacific climate and the LC eddy shedding process.
Previous studies on the teleconnection between the Atlantic and the Pacific showed that El Niño, a Pacific event, can have a strong impact on wind and circulation in the Atlantic (Enfield and Mayer 1997; Alexander and Scott 2002; Kennedy et al. 2007; Smith et al. 2007). The frequent swing in the trade winds and resulting WSCs in the Atlantic may favour more eddy shedding in the GoM (Chang and Oey 2013b). Detailed processes determining how the basin-scale teleconnection influences LC eddy shedding clearly need further study that combines observations and numerical model sensitivity experiments.

5. Summary
Three patterns and corresponding temporal evolution of LC SSH were extracted from 21 years of weekly satellite SSH data using the SOM method. In most cases, the LC evolution follows a normal–extension–retraction cycle. Transitions from normal pattern (P1) to extension pattern (P2) and from extension pattern (P2) to retraction pattern (P3) occur in June and the end of August, respectively.
The weekly FO analysis of the LC retraction pattern (P3) indicates CS WSC has a major influence on LC eddy shedding. The increase of CS WSC from June to November favours the LC eddy shedding at higher frequency during that period. On the interannual time scale, the significant relationship between ONI and AFO of P3 suggests a possible connection between Pacific climate and LC eddy shedding frequency, which needs further study. Due to the fully three-dimensional nature of the LC and its high nonlinearity (e.g., Lugo-Fernández 2007), realistic dynamical modelling study is needed to better understand causes of LC shedding, vertical structure of circulation, as well as their responses to various forcing agents and climate signals.

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