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The influence of 2015-16 El Niño on the record-breaking mangrove dieback along northern Australia coast

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ABSTRACT

This study investigates the underlying climate processes behind the largest recorded mangrove dieback event along the Gulf of Carpentaria coast in northern Australia in late 2015. Capitalizing on the satellite observation-based mangrove green-fraction dataset, variation of the mangroves during recent decades are studied, including their dieback during 2015. The relationship between mangrove greenness and the climate conditions is examined using available observations and by exploring the possible role of the mega 2015-16 El Niño in altering the favorable conditions for the mangroves. The mangrove greenness is shown to be coherent with the low-frequency component of sea-level height variation related to the El Niño southern oscillation (ENSO) cycle in the equatorial Pacific. The sea-level drop associated with the 2015-16 El Niño is identified to be the crucial factor leading to the dieback event. A stronger sea-level drop occurred during austral autumn and winter, when the anomalies were more than 12% greater than the previous very strong El Niño events. The persistent drop in sea-level height occurred in the dry season of the year when sea-level was seasonally at its lowest, so potentially exposed the mangroves to unprecedented hostile conditions. The influence of other key climate factors is also discussed, and a multiple linear regression model is developed to understand the combined role of the important climate variables on the mangrove greenness variation.

1 Introduction

In late 2015, about 8,000 hectares of the forested tidal wetland along a 1000 km stretch of the coastline of Australia's Gulf of Carpentaria (see figure 1a for its location) experienced an extensive dieback of its mangroves¹⁻³. Mangroves are an important part of the local ecosystem, providing habitat for many marine and coastal species, protecting coastlines from extreme weather and erosion, filtering out sediment from river run-off to protect seagrass, as well as absorbing substantial amounts of greenhouse gases⁴⁻⁶. Because of their ecological importance, considerable attention has been paid to this dieback event to understand its extent, cause, and severity⁷⁻⁹. Mangrove dieback events were also observed during 2015-2016 in Kakadu National Park, Northern Territory¹⁰ and in a semi-arid stand near Exmouth, Western Australia¹¹, but those events were not as extensive and severe as the Gulf of Carpentaria mangrove dieback. Preliminary observations have ruled out any direct influence of human activities (e.g., oil spills), pathogens or any extreme weather event (e.g., tropical cyclone) at this time. Thus the prevailing climate conditions, possibly driven by the mega-El Niño that occurred during 2015-16, might be the primary factor for the dieback^{1,12}, but the precise connection with the climate drivers has not previously been identified.

Mangroves exhibit wide physiological tolerance and can intake fresh groundwater as well as saline water during tidal inundation^{13,14}, thus they can withstand a wide range of daily and seasonal tidal variations. However, a major departure from the usual hydrological regime can induce physiological stress to the mangroves and their spatial extent may be considerably modified^{11,15}. Earlier studies have suggested a range of climatic factors that might be important in causing stress to the mangroves, including unusually high temperatures, below-normal rainfall and humidity, and sustained low sea-level^{1,12}. Harris et al.¹⁶ described the 2015 event as a climatic 'press-pulse' process, with a long-term, slow climatic press changing the climatic envelope ecosystems are adapted to, with the impact of this press amplifying the extreme of pulse events. Overlaid on an ongoing warming trend associated with global warming, It is suggested that the 'pulse' was not sudden for the Gulf mangrove dieback but comprised a sustained months-long drop in sea-level due to the 2015-16 El Niño¹², and enhanced air temperatures with persisting drought condition acting to limit freshwater input to the coastline. Owing to their shallow roots,

two mangrove species, particularly *Avicennia marina* and to a lesser extent *Rhizophora stylosa* are vulnerable to prolonged below normal sea-levels. Sippo et al.¹⁷ concluded that dry conditions combined with porewaters enriched in iron associated with unprecedented below-normal sea levels were the potential causes of the dieback. As a result of months-long episode or permanent sea level drop, a change in the ecosystem can be triggered, with a shift from mangrove forest to drier and more saline saltmarsh, where only a few specialised plants can only grow¹⁸.

Sea level drop is possibly a major issue in the dry season, when this stress factor is compounded by persistent low rainfall¹⁹. However, mangroves are reasonably heat tolerant²⁰ and the Gulf mangroves routinely cope with the dry season every year. Furthermore, they have apparently remained mostly unaffected during other El Niño events in the recent past, even during previous strong El Niños such as 1997-98¹. This evokes the questions: what was unusual about the 2015-16 El Niño that resulted in the massive dieback? Was it simply its magnitude, or was the timing and possibly co-occurrence with other climatic factors crucial? Any effort to address these outstanding questions will help in understanding how this environmental catastrophe occurred and should facilitate improved capability for monitoring and prediction, which promises great value in the optimization of risk management for policymakers, and communities in the future.

In this study, we examine the climate conditions during the 2015-16 El Niño in the context of conditions during previous El Niño events. We utilise satellite-based mangrove greenness observations to investigate the greenness changes associated with variations of observed sea level height (SLH), rainfall, surface air temperature, and a measure of vegetation moisture stress formed by comparing potential and actual evapotranspiration. Based on the observed mangrove variations during previous El Niño events and any associated droughts and heat extremes, we provide a plausible explanation for the 2015 dieback that points at the un-fortuitous seasonal timing and magnitude of the 2015-16 mega El Niño as the primary culprit. We also derive a statistical relationship between mangrove changes and SLH, temperature and evapotranspiration variations to provide a stress index model that can be used to anticipate variations of mangroves associated with variations of climate conditions in the future.

2 Data & Method

The mangroves of the Gulf of Carpentaria coast typically grow in a narrow 100-200m span along the shoreline, and thus require high-resolution data for effective monitoring. The satellite-based Landsat Thematic Mapper (TM), enhanced TM and operational land imager provide a reasonably high spatial resolution (30m) observation of mangrove greenness every 16 days (cloud permitting) since late 1980s²¹. The vegetation sampling sites are randomly positioned within the maximum historical extent of the mangroves (figure 1a), as mapped by the Mangrove canopy cover version 2.0.2²². Plots, 90 × 90 m, are stratified to represent three distinct mangrove communities: estuarine, hinterland, tidal, and three dieback classes based on the green-fraction reduction during 2015-16: high (80-100%), moderate (60-79%), and low (30-59%). Dieback severity is classified based on relative cover loss between pre-dieback (preceding dry season of 2015) to post-dieback (following wet season of 2015-16). All sites are spaced a minimum of 500m apart unless areas are of different severity classes. Fractional cover products are derived by separating pixels into constituent fractional exposed greenness cover using spectral unmixing algorithms with joint remote sensing research program²³. We extract fractional cover scenes by removing pixels influenced by cloud, cloud shadow, and surface water, using bitwise masking “Water Observation Feature Layers” derived from the Water Observation from Space²⁴. A mean monthly green-fraction is derived for each site from all available fractional greenness coverage observations from 1987 to 2020. The outliers are excluded from the time series using a rolling mean to detect data points that exceeded more than 20% cover change from the three-month average. All fractional cover products are extracted using Digital Earth Australia’s sandbox platform²⁵, and the products accessed from the open data cube²⁶.

SLH variations are monitored with observations from the Australian Bureau of Meteorology (BOM) Milner Bay, Groote Eylandt tide-gauge station (location shown in figure 1a) during 1993-2018. As the seasonal cycles of the SLH are similar across all the stations in Gulf of Carpentaria^{27,28}, it is reasonable to use a single tide-gauge location as representative of SLH in the Gulf of Carpentaria. We also utilize monthly Bluelink ReANalysis (BRAN) SLH analyses²⁹, which are available globally between 75°S – 75°N at 10km resolution. BRAN assimilates observations of SLH (based on satellite altimetry and in situ observations) together with in situ observations of temperature and salinity into a global ocean model using ensemble optimal interpolation of observational data assimilation system to provide several 3-dimensional time-varying analyses of ocean temperature, salinity, currents and SLH.

Monthly rainfall and maximum surface air temperature (T_{max}) dataset on 0.05° grid are obtained from the Australian Water Availability Project (AWAP)³⁰. This data is available across continental Australia back to 1911. We use monthly surface zonal wind analyses from European Centre for Medium-Range Weather Forecasts (ECMWF) Interim reanalysis (ERA-I)³¹, which are available globally on a 1.5° grid during 1979-Aug 2019. An evaporative stress index (ESI)^{32,33} is used to examine the evaporative stress along the Gulf’s coast. This index is the normalized ratio of evapotranspiration to potential evapotranspiration, which are derived from the outputs of the AWRA-L landscape water balance model over Australia (version 6) and is indicative of surface moisture supply and evaporative demand³⁴. The rainfall, T_{max} and ESI data are averaged over coastal Gulf region (shown as red box in figure 1a).

The anomalies are calculated by removing the seasonal cycle (time mean and first three harmonics of climatological annual cycle) from the variable and the linear trends are removed from the anomalous timeseries. The anomalies are standardised by their own standard deviation to obtain the normalised anomalies. We identify previous strong El Niño events as when the oceanic Niño3.4 index exceeds 1.5 standard deviations in the November-January season, when El Niño conditions typically peak in central to the eastern equatorial Pacific. This Niño3.4 index is an average of sea surface temperature (SST) over the central Pacific region ($5^{\circ}S - 5^{\circ}N$, $170^{\circ}W - 120^{\circ}W$), calculated using HadISST dataset³⁵. To increase the sample size of strong El Niño events, we assume linearity of La Niña events (being opposite to El Niño) and include in our composite of the previous strong La Niña events but flip the sign of the anomalies. Based on this index, we identify the austral summers of 1957-58, 1965-66, 1972-73, 1982-83, 1987-88, 1991-92, 1997-98, 2015-16 as strong El Niño years and 1973-74, 1975-76, 1988-89, 1998-99, 1999-2000, 2007-08, 2010-11 as strong La Niña years. The Interdecadal Pacific Oscillation (IPO) is monitored using SST-based IPO tripole index (TPI)³⁶.

3 Results

Historical perspective of the mangrove green-fraction variability

The monthly timeseries of the mangrove greenness in the three dieback categories for over 30 years are displayed in figure 1b-d. The green-fraction dataset is averaged over all the sites in each category and the mean monthly green-fraction timeseries with its range across the sites are shown. A substantial drop in green-fraction is evident for the moderate and high dieback categories during 2015, with some locations in the high category experiencing near 100% loss of greenness. However, abundant variability is also evident, primarily associated with the seasonal cycle (greenness peaks in austral autumn), but also with other low-frequency variations. For instance, the green-fraction has a sustained 20% reduction around 1994 and subsequent recovery during the late 1990s. However, the magnitude of the drop during 2015 is unprecedented at least in the last few decades (e.g., Duke et al.¹). The seasonal variation of the mangrove greenness is $\sim 10\%$ of the mean, with the maxima occurring in March-April and the minima around October (figure S1). During 2015, the seasonal reduction of green-fraction started about a month or two before the climatological minima and it dropped noticeably for moderate and high categories by Aug-Sep.

Impact of El Niño events

Considering the coincidence of the strong El Niño of 2015-16 with the 2015 dieback event, we examine the variation of different climate variables during the 2015-16 El Niño and compare them with previous strong El Niño conditions. This provides a paradigm for understanding the extreme climate conditions that might have stressed the mangroves during 2015-16. Figure 2 shows the evolution of the normalized Niño3.4 index, mangrove green-fraction, T_{max} , rainfall, ESI along the Gulf coast and SLH anomalies at Milner Bay from the development period of the strong El Niño events (the preceding February) to their demise (the following April). The influence of strong El Niños on the variables are obtained by lag-regressing the normalised anomalies onto December Niño3.4 index and the derived regression coefficients are scaled by the normalised Niño3.4 index magnitude (1.44) during strong ENSO. SST in the Niño3.4 region of the eastern equatorial Pacific tends to peak in December. Mangrove green-fraction reduces after the peak of El Niño events, but this reduction in greenness is not usually large. During a typical El Niño event, T_{max} in the Gulf region is near to neutral in the development stage and then becomes higher than normal at the peak of El Niño (i.e., during December-February). For this Gulf region, rainfall displays little variation during El Niño, however evaporative stress is typically stronger than normal (i.e., negative anomalies) throughout the course of El Niño. Finally, SLH in the Gulf is also seen to be typically lower than normal from late winter through to early autumn, consistent with the strong influence of the Pacific on the Gulf SLH variation as revealed in Oliver and Thompson²⁸.

During the 2015-16 El Niño, the Niño3.4 SST index was 50% stronger at its peak in December 2015 compared to the composite. It started to become positive earlier since the preceding austral winter and lasted later into the following autumn. A strong positive T_{max} anomaly developed around April-May, but it reversed sign during the peak of the event in December 2015. Lower than normal rainfall accompanied the high T_{max} in the preceding winter, but rainfall was near normal during the peak of the event during spring and early summer 2015, consistent with the weak rainfall anomalies observed during previous strong El Niño events. And, in contrast to other El Niño events, ESI was very negative in the autumn, winter and spring of 2015, indicating persistence of a stressful environment for the mangroves. Lastly, SLH became much lower than normal 4-5 months earlier than is typically observed during El Niño and remained lower longer into the following autumn. The de-trended SLH time-series at Milner Bay indicates that the drop during 2015 was unprecedented. From this cursory analysis, the extreme drop in SLH together with the preceding strong increase in negative ESI and high T_{max} during the preceding winter could have provided the necessary stress to cause the extreme dieback that was sustained during 2015 (figure 2). The persistent below-normal SLH especially during hot and dry conditions would have resulted in limited tidal mixing and that might provide a stressful condition for the mangroves, which generally intake water during tidal inundation¹⁴.

Connections between mangrove greenness and climate variables

We further explore the historical relationship between SLH and mangrove greenness and their association with El Niño conditions in the equatorial Pacific using cross-spectral analysis between SLH and green-fraction. This analysis is useful for quantifying the relationship between two variables as a function of frequency and estimating their phase relationship. We compute the coherence squared by Fourier transforming the deseasonalized monthly times series for 1994-2016, which provides a band width of $1/276\text{-month}^{-1}$. The coherence squared is formed after smoothing the raw power and cross power estimates with a 9-point box car, yielding an effective bandwidth of $9/276\text{ month}^{-1}$.

The top panels of figure 3 show the coherence squared and phase lag between the monthly SLH at Milner Bay and the index of mangrove green-fraction for three dieback categories. The cross-spectra are found to be largely similar for all three categories - the peak coherence up to 0.45 occurs around 0.03-month^{-1} frequency (period ~ 2.8 years) and are broadly high for frequencies lower than $1/25\text{ month}^{-1}$ (period ~ 2.1 years). There are no other significant spectral peaks at higher frequencies. The phase lag (about 45°) indicates that low frequency (period longer than 2 years) greenness variation lags SLH by about 5-6 months, which is reasonable as it should take a few months for the mangroves to react to lower or higher than normal sustained SLH change.

A similar quantification of the relationship between Gulf's mangrove green-fraction and coastal Gulf region area averaged ESI is shown at the bottom panels of figure 3. A weaker spectral peak at periods longer than 2 years is also evident with about 3-4 months phase lag. We also examine the coherence-squared between green-fraction and T_{max} , and rainfall (figure S2) and unlike SLH and ESI, these variables generally show no significant coherence with greenness. Only the peak coherence between T_{max} and green-fraction for moderate and high dieback categories at the frequencies lower than 24 months is found to be marginally significant ($p=10\%$). An out-of-phase relationship is also noted between green-fraction and T_{max} , indicating reduction of green-fraction due to low frequency warming over the Gulf.

More insight into the cause of the low frequency SLH anomalies associated with greenness variations is gained by examining the spatial distribution of the coherence-squared between the gridded BRAN SLH and mangrove green-fraction index anomalies. The coherence-squared and phase lag for the dominant low frequency band around the 30-month cycle are shown in figure 4a. Here we form the index of greenness by averaging standardized greenness anomalies. Mangrove greenness at Gulf of Carpentaria sites is coherent with SLH throughout the Gulf and on the north-west coast of Australia, in the seas of the Indonesian archipelago and in the north-west Pacific, with a constant phase lag of about $1/8$ cycle (~ 4 months). The greenness is also found to be coherent with SLH on the equatorial central and eastern Pacific, with a distinctive horse-shoe pattern. The phase is roughly 180° shifted from that in the Gulf and surrounding seas, indicating that greenness is coherent with an El Niño-like fluctuation of SLH (i.e., low SLH in the Gulf and western Pacific coincides with high SLH in the central and eastern Pacific). However, the pattern of coherence in the central and eastern Pacific, with its distinctive horseshoe shape, is more indicative of the interdecadal variation of ENSO referred to as the IPO (e.g., Power et al.³⁷). This sensitivity of greenness to the lower frequency component of ENSO presumably arises for two reasons. First, the lower frequency component of ENSO drives a stronger variation of SLH in the Gulf than it does by the typical interannually varying El Niño event. This is confirmed by examining the coherence of SLH in the Gulf with SLH at every grid point for the dominant frequency band (figure 4b). The peak coherence with SLH in the eastern Pacific, which has the opposite phase to that in the Gulf, occurs just east of the dateline and with a relative minimum in coherence along the equator in the eastern Pacific. This is a typical IPO feature and the coherence between the IPO index and SLH confirms this structure (figure 4c). Secondly, as greenness is effectively a slowly varying parameter, it is more coherent with the lower frequency components of El Niño. These two processes together result in a pattern of coherence between greenness and SLH that looks more like the IPO than it does El Niño. That is, greenness is more responsive to the low-frequency tail of ENSO variability than it is to the higher frequency interannual components.

The overall pattern of coherence of SLH with greenness thus reflects the spatial structure of ENSO, suggesting that the El Niño driven SLH variability is linked with mangroves. During El Niño, westerly wind anomalies in the western Pacific act to elevate the SLH to the east with a Kelvin wave structure and lower SLH in the western Pacific with Rossby wave structure³⁸. The lower SLH anomalies in the west Pacific travel through the Indonesian seas and act to lower the SLH in the Gulf²⁸ and down the west coast of Australia where it is transmitted as a coastally trapped Kelvin wave³⁹. The broader meridional structure of the SLH variation in the central and eastern Pacific is coherent with Gulf greenness and that reflects the structure of the lower frequency tail of El Niño, and it is associated with slower westward propagating Rossby waves off the equator in the eastern and central Pacific. However, it remains in question why the mangroves dramatically declined in 2015 but survived during other El Niño events. We explore this outstanding question by examining SLH variation during strong El Niños next.

Sea level variation during strong El Niños – how did 2015-16 differ?

To gain insight into the SLH variation during El Niños, we look at the timing of the SLH decline during 2015-16 and compare that with other strong El Niño events. Figure 5a displays the annual cycle of SLH in Milner Bay along with the variations during 2015-16 and two other strong El Niño events (1982-83 and 1997-98). The SLH maximum occurs during austral summer

and the minimum is noted during late winter. Forbes and Church²⁷ explained seasonal SLH variation as a result of the seasonal cycle of wind stress over the region (i.e., westerlies during the summer monsoon acting to raise SLH and peak easterly trade winds during winter that act to lower SLH). The three El Niño events shown in figure 5a result in lower-than-normal SLH from about April of the year El Niño commences through to the following March. But it is noted that no severe mangrove dieback was recorded during the 1997-98 El Niño event, which has been referred to as “El Niño of the century” for its extraordinary magnitude and influence on the global weather and climate⁴⁰. The most dramatically different behaviour of the 2015-16 El Niño is the much stronger negative SLH anomalies during austral autumn and winter, when the anomalies were more than 12% stronger than the previous two very strong El Niño events. The extreme sea level decline in 2015 coincided with the seasonal minimum, so potentially exposed the mangroves to unprecedented low SLH.

The extremity of the low SLH during 2015 is quantified by computing the cumulative deficit of detrended SLH. We do this both for the cumulative anomaly and for the cumulative anomaly below the climatological minimum, which is assumed to be the threshold below which the mangroves cannot be sustained. We show this for the 1997-98 (figure 5b) and 2015-16 (figure 5c) El Niño events. This analysis indicates that cumulative stress due to below-normal SLH was double in 2015-16 compared to 1997-98. Sustained SLH below the climatological minimum continued into October 2015 but was confined to only moderate values during June-August 1997.

We further evaluate the SLH and surface zonal wind evolution around the equatorial Pacific during the strong El Niño events of 1997-98 and 2015-16 (figure 6). This domain was chosen as the onset of the wind anomalies occurs along the equator⁴¹ and the wind-forced SLH changes in the equatorial western Pacific appear to have a strong relationship with the Gulf’s SLH variation. Figure 6 indicates that the drop in SLH around the equatorial Pacific during peak El Niño months of 1997-98 were stronger relative to the same period of 2015-16. But during early and middle months of 2015, the SLH drop was greater compared to the same period in 1997. The below-normal SLH anomalies during 2015 can be traced back to mid-2014, when an El Niño event false-started⁴². As the westerly zonal wind anomalies strengthened around late austral autumn of 2015, the wind-induced downwelling Kelvin wave acted to lower SLH in the western Pacific for a prolonged period in early and middle months of 2015, especially in the dry season of the year.

Our analysis suggests that the low-frequency (periods longer than 2-3 years) SLH is the key driver of the mangrove dieback event. A persistent drop in SLH associated with El Niño likely establishes a stressful environment for the mangroves, especially during the dry months of the year (late austral autumn and winter). Additionally, evaporative stress, as quantified by the ESI, appears to be important for the mangroves as does T_{max} . The 2015 El Niño was thus catastrophic because SLH dropped considerably much earlier than usually occurs during El Niño. It exposed the mangroves to a sustained period of SLH well below its climatological minimum, together with atmospheric anomalies that resulted in increased evaporative stress during the preceding autumn and winter seasons. This sustained low relative sea-level was captured by the mangrove stress index defined in Duke et al.³. In this study, we go further and include other potential factors to derive a mangrove stress index.

Mangrove stress index

To understand the combined role of the potential climate variables on the dieback event, we reconstruct the mangrove green-fractions for each of the dieback categories using SLH, ESI and T_{max} anomalies. A multiple linear regression model is developed to estimate the greenness variations with seasonal cycle removed, detrended, standardised 6-month lagged SLH at Milner Bay, 3-month lagged ESI, and T_{max} anomaly time-series over the coastal region of the Gulf as the predictors. A 12-monthly running mean is applied to the predictors to eliminate the high-frequency variability. The mathematical model for the multilinear regression analysis is as follows:

$$Greenfraction_{estimated} = b_0 + b_1.SLH + b_2.ESI + b_3.T_{max} \quad (1)$$

The estimated and observed green-fractions for three dieback categories over the period of 1994-2018 are displayed in figure 7 and the associated regression coefficients with relevant statistics are given in Table 1. Consistent with our previous results, SLH has the dominant contribution, and the associated coefficients are significant at the 5% level for all three categories. By contrast, temperature and ESI show weaker association and the regression coefficients for the T_{max} anomalies are not significant at the same confidence level.

The correlations between estimated and observed green-fractions range between 0.47 and 0.62. Relatively higher correlations are noted for the high dieback, suggesting adequate skill of the regression model in predicting the vulnerable mangroves. The reconstructed green-fractions also show an unprecedented reduction in greenness during 2015, though the estimated decrease is much weaker than observed. Nevertheless, the analysis in figure 7 confirms that the multilinear regression model can be used to construct the mangrove stress index for estimating the future condition of the mangroves using dynamical seasonal forecasts (e.g., Long et al.⁴³) and climate model projections, and it will form the focus of future work.

Table 1. Regression coefficients for three dieback categories during 1994-2018 from the multiple linear regression model (1). Regression coefficients significant at 5% level are marked in bold.

Dieback class	Regression Coefficients			
	b_0	b_1	b_2	b_3
Low	0.151	0.427	0.386	-0.076
Moderate	0.377	0.606	0.12	-0.089
High	0.216	0.604	0.325	-0.087

4 Discussion

We investigate the climate conditions during the 2015 mangrove dieback event along the southern coastline of the Gulf of Carpentaria using available observations and reanalysis datasets. The decline in SLH associated with the positive phase of ENSO is shown to be primarily responsible for the dieback event. The eastward shift of the equatorial Walker circulation during El Niño drives a dynamical drop in sea level throughout the Gulf of Carpentaria, which exposes the mangroves to hostile conditions. Our analysis over the period 1994-2018 suggests that low-frequency SLH variability (i.e. a period longer than 2 years) associated with ENSO explains about 50% variation of the observed low frequency variation of mangrove green-fractions. Slowly varying greenness is shown to be more strongly associated with the lower frequency (decadal) component of ENSO than its higher frequency interannual components.

The phase lag between SLH change and the variation of mangrove greenness (greenness declines after SLH drops) indicates that green-fraction reduces about 5-6 months after the persistent drop in SLH. During 2015, a much stronger SLH drop, about 12% greater than previous strong El Niño events, and much earlier onset of the decline (SLH dropped in June-July 2015 whereas typical SLH drop during El Niño does not occur until Aug-Sep) occurred during June-July when SLH is seasonally at its lowest, and that potentially exposed the mangroves to unprecedented conditions. A prolonged period with low SLH along with stronger dry and warmer conditions during a typical El Niño provided a stressful environment for the mangroves. Such hostile conditions during 2015 caused canopy loss of the mangroves and eventually led to the massive dieback event.

The drop in SLH in the Gulf in 2015 was also associated with a drop in SLH along the western Australia coast, which typically occurs during El Niño. This decline in SLH was also responsible for a relatively small-scale mangrove dieback event along the Western Australia coast^{11,44}, reflecting the rapid adjustment of SLH along the north and west coasts of Australia in response to El Niño (e.g., Potemra⁴⁵). The below normal SLH that began in 2014 due to weak El Niño condition in that year were further lowered in early and middle months of 2015 by the strong westerly wind anomalies around the equatorial western Pacific leading to critically below-normal SLH in the Gulf of Carpentaria during the dry season of 2015. The dynamical mechanism for the remote response of sea level during El Niño in the Gulf (e.g. Oliver and Thompson²⁸), along the western Australia coast is well established and what set the 2015 El Niño apart from previous events, in addition to its strong amplitude, was its early onset.

The severity of the 2015 mangrove dieback in the Gulf of Carpentaria is found to differ with geographical location, primarily due to distance to the shoreline, local elevation, and distance from the creek^{2,3}. While the previous existence of the mangroves has been noted at a few of the locations in our study, mangroves have apparently emerged since the mid-90s at many of the locations, perhaps in response to the rising SLH. There is a detectable impact of the 1997-98 El Niño on the green-fraction at a few sites, causing ~20-40% loss of greenness of the mangroves. In contrast, some of the plots are almost unaffected in the last few decades, even during the 2015-16 dieback event. Some of those plots that have not sustained extensive dieback in 2015, is a frontal stand dominated by *Rhizophora stylosa*, while other unaffected plots are in a relatively broad and lush region next to a tidal creek. The rest of the dieback plots are at least a few meters (60-200 m) away from the shoreline and having little elevation from the mean SLH. These contrasting patterns of dieback at fine-scales emphasise the importance of hydro-geomorphic setting (height above mean SLH, distance to drainage channels) and hydroperiod (duration of tidal inundation) as important factors that determine the severity of the dieback. An assessment of the plot-scale geomorphic variation is beyond the scope of this present study.

It may be inferred from our analyses and assessment of Duke et al.¹ that the 2015 mangrove dieback event was primarily driven by large-scale stressors (e.g., decadal ENSO variability), however, regional variation of mangrove species, inundation period, distance to drainage channels influenced extent of the dieback. This motivates further investigation using high-resolution green-fraction dataset across the dieback locations to explore the small-scale variation of the dieback event. As the global

climate warms⁴⁶, and features of ENSO may change⁴⁷, mangroves are likely to experience further stress (e.g. Sippo et al¹⁷). Future investigation could include an examination of changes in the large-scale climate stressors in climate model projections to determine if there is an increased likelihood of experiencing mangrove stress events similar to 2015 in the Gulf of Carpentaria in a warming climate.

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Author contributions statement

All authors contributed to discussions and writing the manuscript. S.A. and H.H.H. conceived the design of the work and S.A. performed the data analysis in discussion with P.H., H.H.H., L.B.H. and J.B., while S.J. and W.D. contributed to process the mangrove green-fraction dataset.

Additional information

Competing interests The authors declare no competing interests.

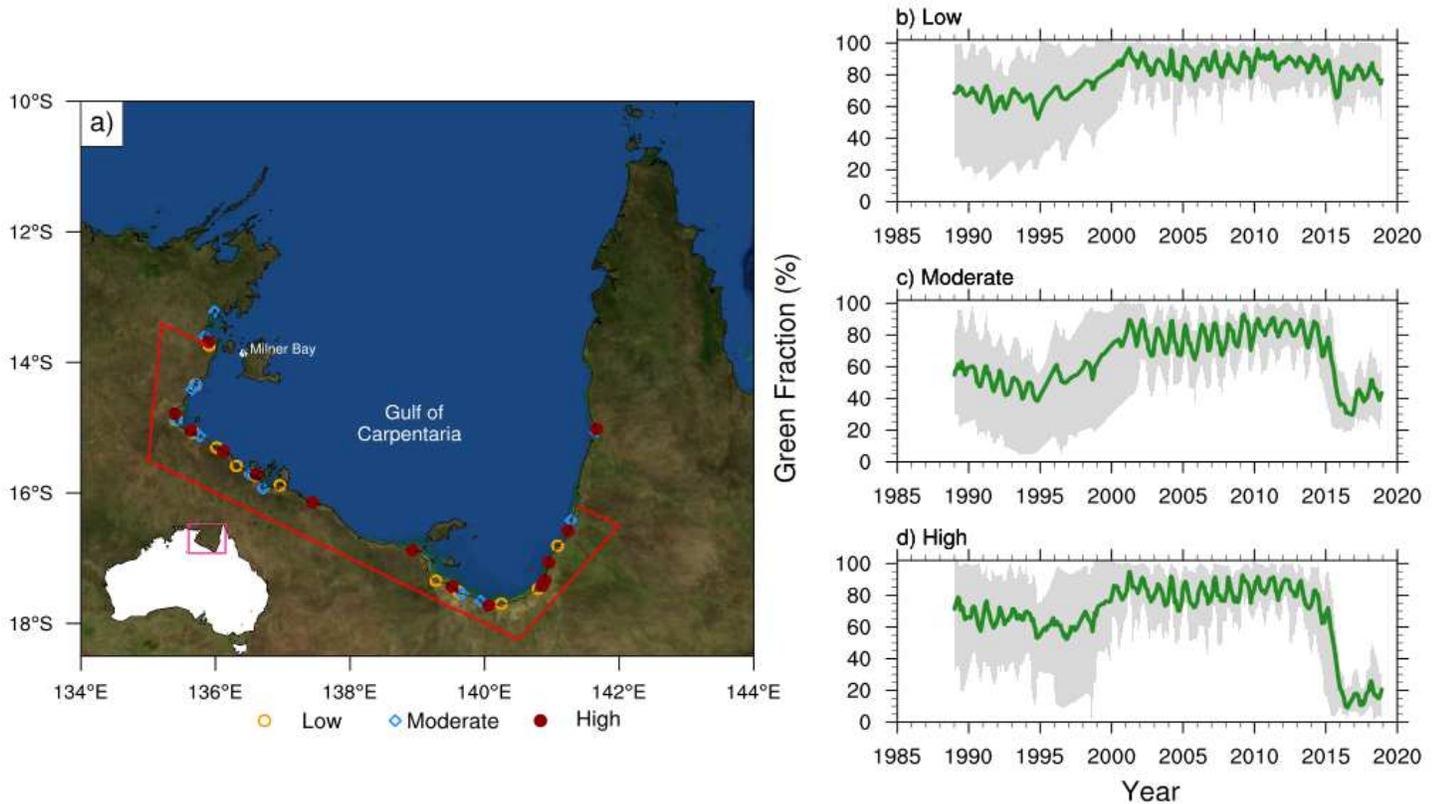


Figure 1. (a) Location of the mangrove green-fraction sampling sites along the Gulf of Carpentaria coastline with an inset in the bottom left showing location of the Gulf of Carpentaria within Australia. The sites with three dieback classes - low (30-59%), moderate (60-79%) and high (80-100%) are marked with yellow circle, blue diamond and maroon dot, respectively. The coastal Gulf region that is used to define T_{max} , rainfall and ESI anomalies are shown in red polygon. (b-d) Time evolution of mangrove green-fractions (in %) for the three dieback categories. The solid green curves indicate the averaged green-fractions across all the sites in each category, while the range of the monthly green-fractions are shown with grey shading.

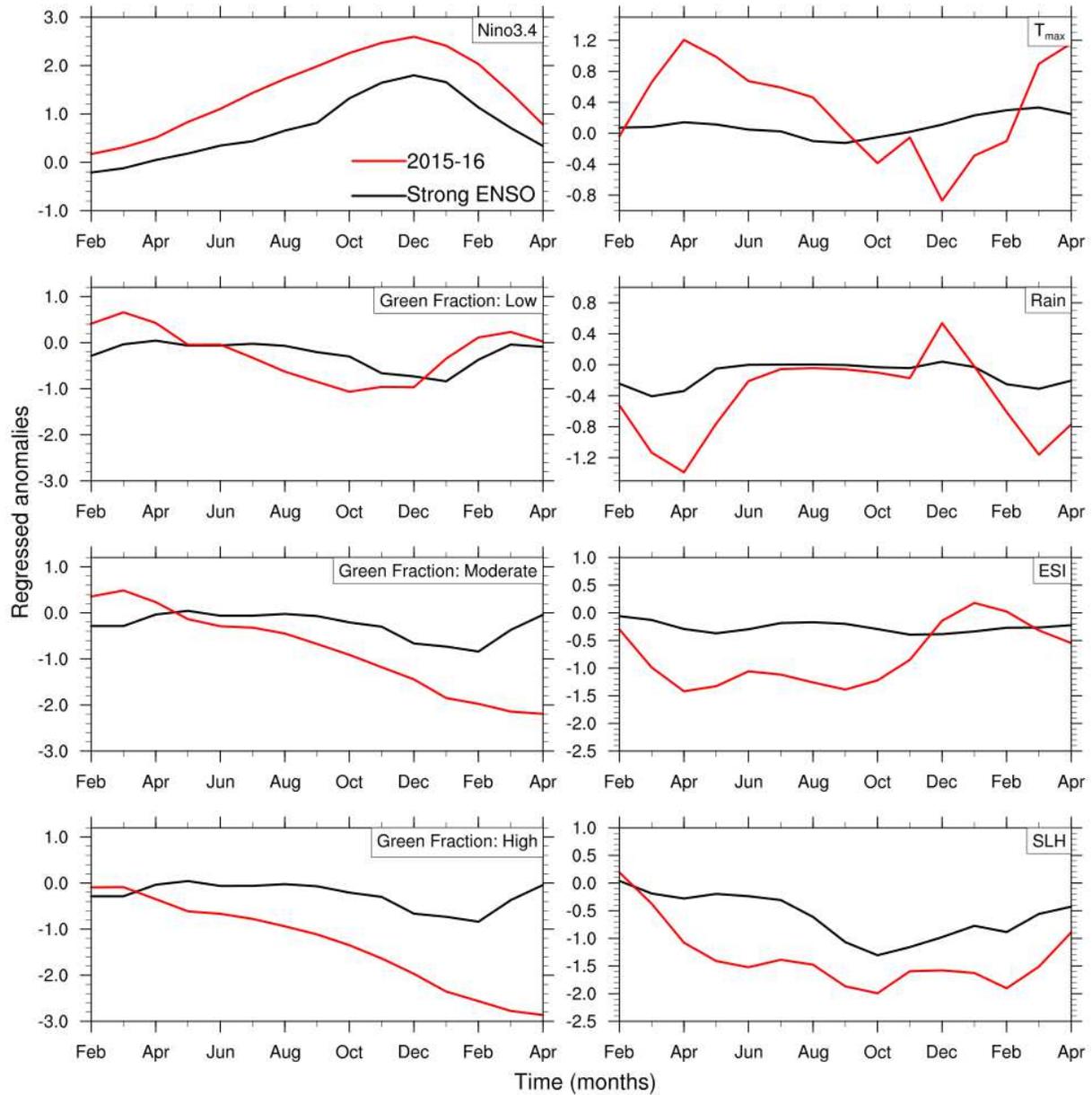


Figure 2. Monthly evolution of Niño3.4, mangrove green-fraction for 3 dieback categories, T_{max} , rainfall, ESI and sea level height (SLH) anomalies during previous strong El Niño conditions (black curve) and 2015-16 (red). All the variables are normalized by their own standard deviation before lag-regressing the normalised anomalies onto Niño3.4 and the regression coefficients are scaled by normalised Niño3.4 index magnitude during strong ENSO.

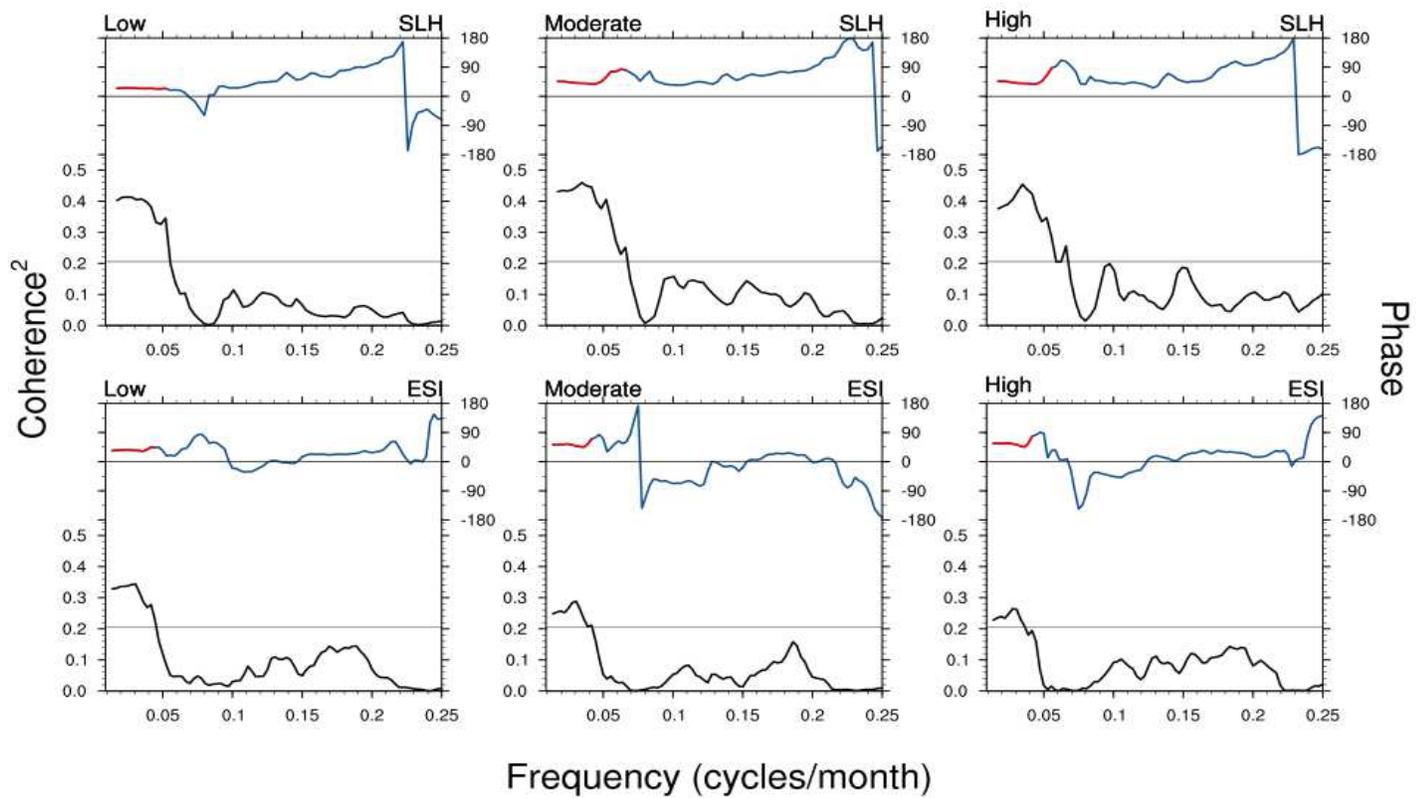


Figure 3. The coherence-squared spectrum (black curve, below) and phase (blue curve, top) between sea-level height (SLH) at Milner Bay and mangrove green-fractions for the three dieback categories of Low, moderate and High (top panel). The same with coastal Gulf area averaged ESI is shown at the bottom panels. The gray-line shows a 5% level of significance for coherence-squared values. Phase lags are shown in red where the coherence squared is significant at 5% level.

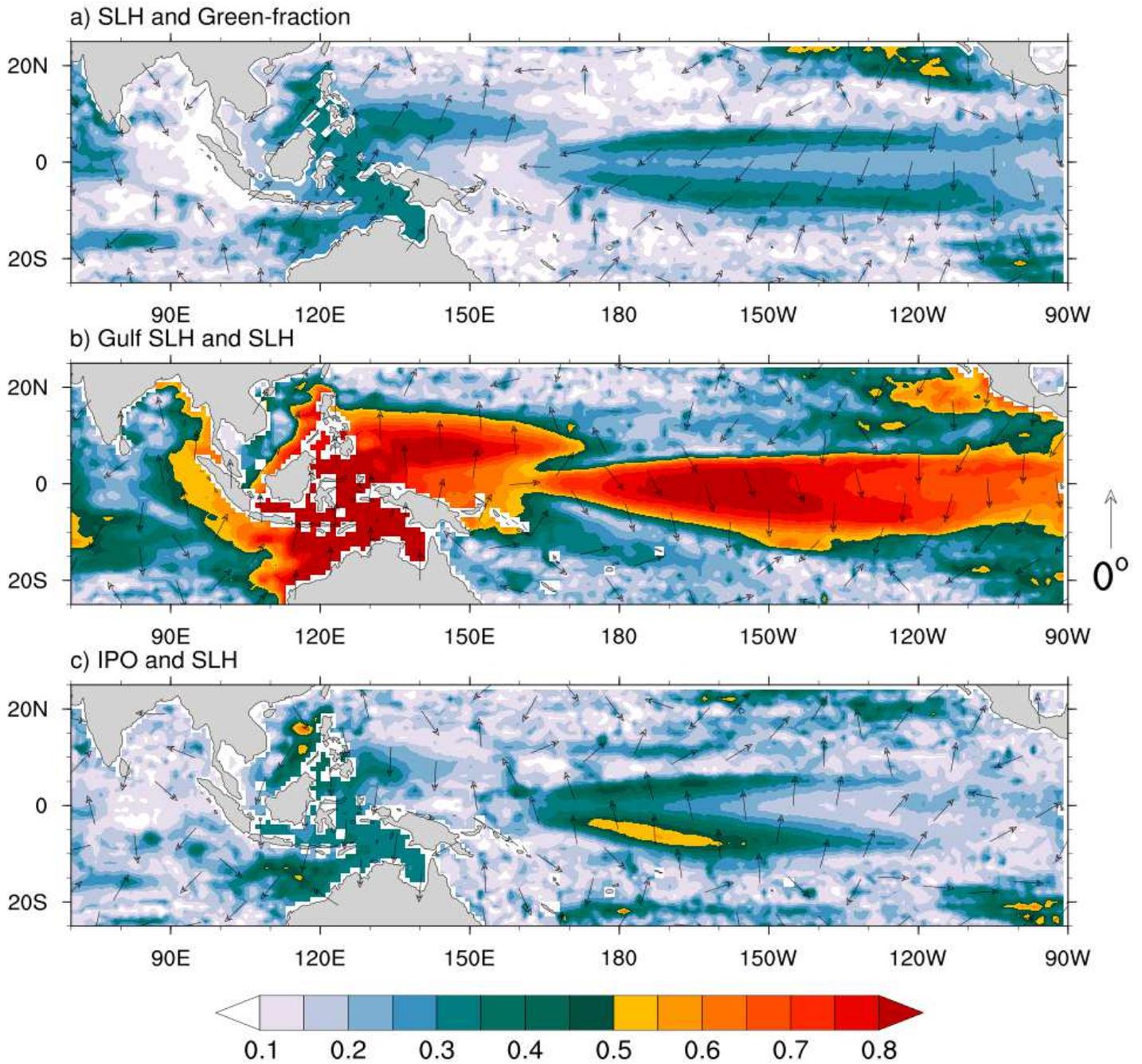


Figure 4. Spatial distribution of the squared coherence at the dominant frequency mode of cross-spectra (shaded) and phase difference (vectors) between (a) sea-level height (SLH) and averaged mangrove green-fraction for three dieback categories. (b) Milner Bay SLH and SLH everywhere, (c) IPO and SLH. Vectors pointing north represent no phase-lag with increasing lag in clockwise direction. The positive lag implies mangrove green-fraction lags SLH anomalies and vice versa.

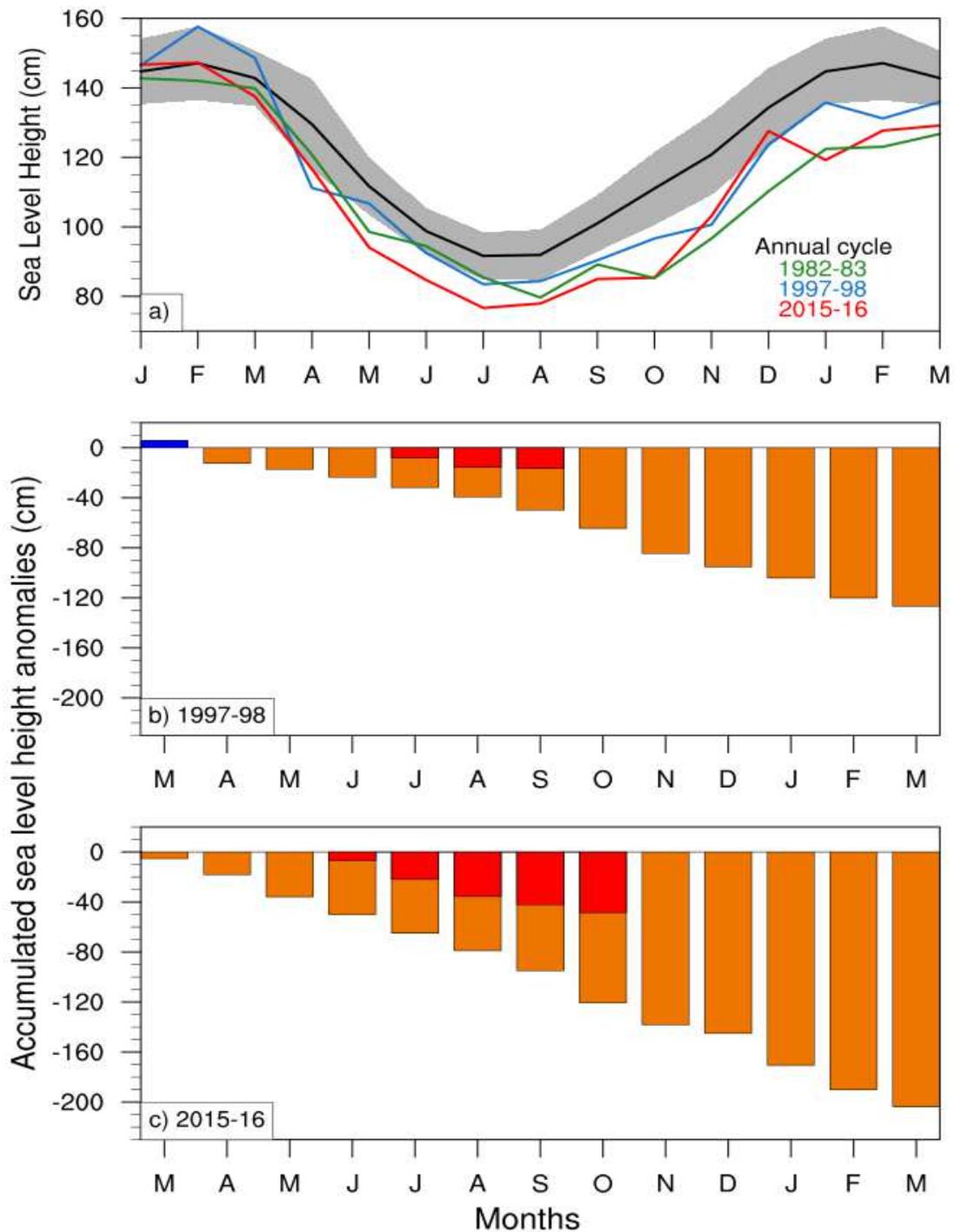


Figure 5. (a) Sea-level height (SLH, in cm) evolution at the Gulf of Carpentaria during Strong El Niño events, along with the climatological annual cycle (black curve) and its two-standard deviation (2σ) range (grey shading). (b-c) Accumulation of drop in SLH anomalies during two recent strong El Niño events are shown as orange bars. The similar accumulation relative to climatological minima are denoted by red bars. The linear trend is removed from the SLH data before the analysis.

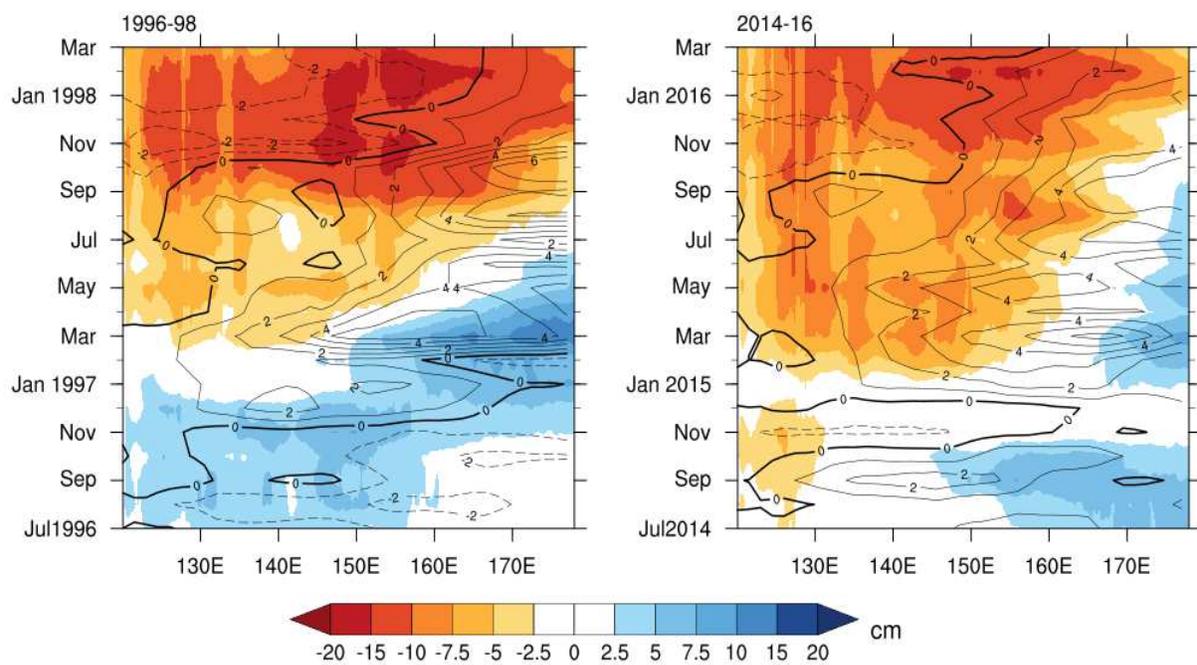


Figure 6. Time-longitude plot of BRAN sea-level height (shaded, cm) and ERA-interim surface zonal wind (contours, $m s^{-1}$) anomalies averaged between $5^{\circ}S$ and $5^{\circ}N$ for two El Niño events during 1996-98 and 2014-16, respectively.

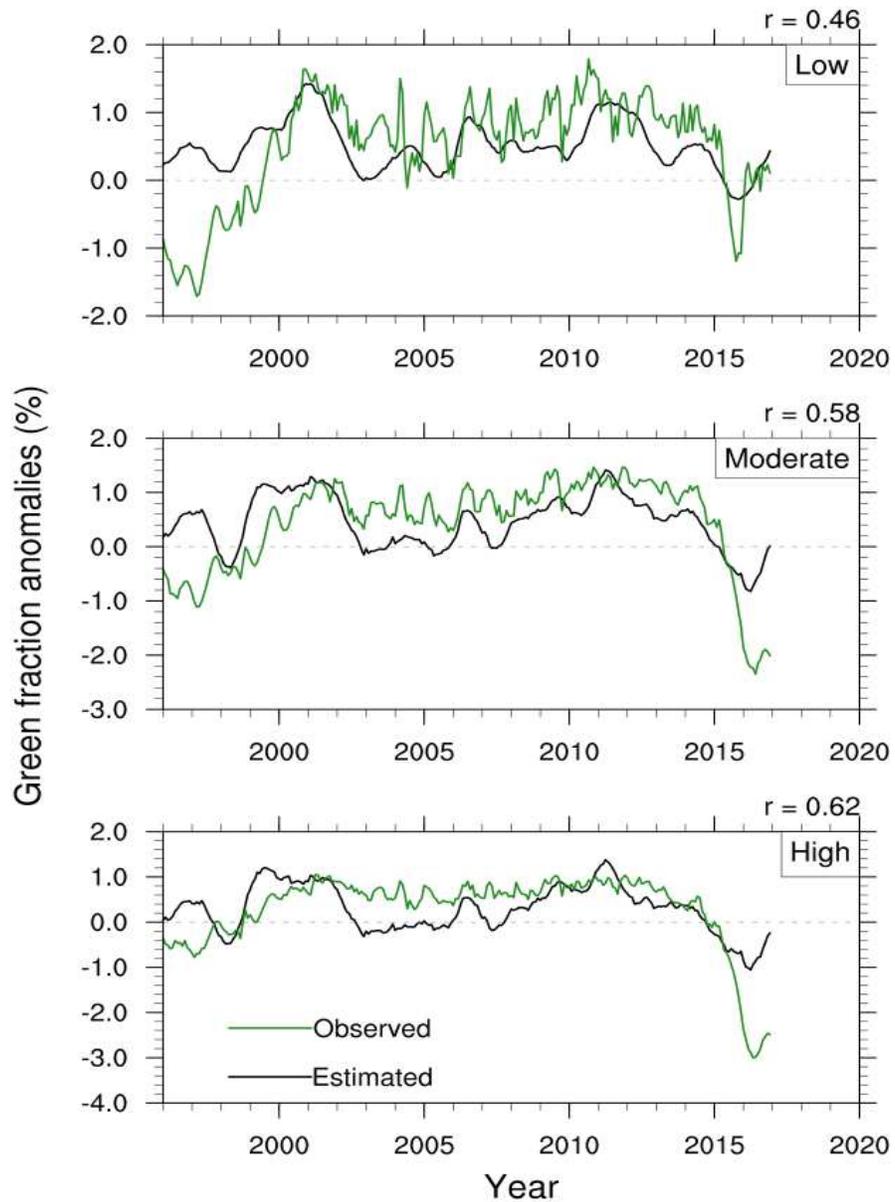


Figure 7. Reconstruction of mangrove green-fraction data using multilinear regression of 6-month lagged sea level height, 3-month lagged ESI and T_{max} averaged over Gulf of Carpentaria coastal region. The predictors are detrended and a 12-monthly running mean is applied to filter out the high-frequency noise. All the datasets are normalised by their own standard deviation. Note that Y-axis ranges are not same in all panels.

Supplementary Files

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