Assessing the drift of Fish Aggregating Devices in the tropical Pacific Ocean

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Abstract. The Tropical Pacific Ocean is characterized by its dominant zonal flow, strong climate dependence on the El Niño Southern Oscillation (ENSO) and abundant tuna stocks. Tuna fisheries in the ~~West~~ Western and Central Pacific Ocean accounted for 55% of world-wide tuna catch in 2019 and are one of the main sources of income in many Pacific island nations. One of the dominant fishing methods in this region relies on the use of drifting Fish Aggregating Devices (dFADs): rafts with long underwater appendages (on average 50m deep) that ~~attract and~~ aggregate fish. Although currents such as the North Equatorial Countercurrent (NECC) and South Equatorial Current (SEC) in the tropical Pacific Ocean vary strongly with ENSO, little is known about the impact of this variability in flow on dFAD dispersion. In this study, virtual Lagrangian particles are tracked for the period 2006 to 2021 over the domain in a 3D hydrodynamic model and are advected in simulations with only surface flow as well as simulations using a depth-averaged horizontal flow over the upper 50 meters. ~~The particle trajectories are used to determine zonal~~, representing virtual dFADs. Zonal displacements, eddy-like behaviour and ENSO variability ~~for drifters that are subjected to either surface or depth-averaged currents~~ are then studied for both types of virtual particles. It was found that virtual particles ~~that are advected by only surface flow~~ advected by surface flow only, are displaced up to 35% farther than ~~those~~ virtual dFADs, subjected to a depth-averaged flow, but no other major differences ~~are~~ were found in dispersion patterns. Strongest correlations between ENSO and virtual dFAD dispersion for the assessed variables were found in the West Pacific Ocean, with Pearson correlation coefficients up to 0.59 for virtual dFAD displacement. Connections between ENSO and eddy-like behaviour were found in the ~~western~~ Western part of the SEC, indicating more circulation and meandering during el Niño. These findings may be useful for improving sustainable deployment strategies during ENSO events, and understanding the ocean processes driving the distribution of dFADs.

1 Introduction

In 2019, tuna fisheries in the ~~West~~ Western and Central Pacific Ocean (WCPO) accounted for 55% of worldwide tuna catch (Williams and Ruaia, 2020). One of the dominant fishing methods, utilized by industrial purse seining fleets, relies on the usage

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of Fish Aggregating Devices (FADs) (Davies et al., 2014). These devices aggregate fish to their location, making it easier to catch large quantities of tuna during one fishing event. These FADs can be grouped in two categories: anchored and drifting FADs (Leroy et al., 2013), where the former are generally used by coastal fleets and artisanal fishers, and the latter (dFADs) are generally used by distant-water and industrial fleets and will be the focus of this research.

In their basic form, dFADs are floating, satellite-tracked bamboo rafts with appendages underwater that can reach to average depths of 50m or more (Fonteneau et al., 2000). Due to these appendages, dFADs move slower than surface drifters (Imzilen et al., 2019). The concept of a dFAD is based around the behavioural tendency of pelagic fish, such as tuna, to associate with floating objects (Fréon and Dagorn, 2000; Castro et al., 2002).

Most technological advances, more often than not kept as trade secrets, share a common goal: increasing dFAD efficiency (Fonteneau et al., 2000). These In the case of dFADs, these advances are essentially a double-edged sword. On one hand, they can help make the fishing more cost-effective and reduce carbon emissions for a purse seining fleet, which is a highly mobile and dynamic fishery. On the other hand however, it could lead to overfishing of stocks, unnecessary by-catch and damage to the ecosystem (Dagorn et al., 2013). Both papers by Dagorn et al. (2013) and Bailey et al. (2013) point out that with proper management these devices can be used as a safe and sustainable fishing method, and building knowledge on dFAD dynamics is important for sustainable management. Similarly, the marine pollution and damage to coastal habitats from abandoned, lost or otherwise discarded dFADs is of increasing concern (Escalle et al., 2019). Here we study these the oceanographic dynamics of dFAD dispersion in the tropical Pacific Ocean, where most tuna is caught within the West Pacific warm pool (Lehodey et al., 1997) and which influences dFAD-tuna interactions (Nooteboom et al., 2023; Pérez et al., 2022).

Presumably, three main surface currents influence dFAD dispersion in the tropical Pacific Ocean: the North Equatorial Current (NEC), the South Equatorial Current (SEC) and the North Equatorial Counter Current (NECC). The first two currents flow eastward under normal conditions and are part of the two basin-wide Pacific circulation gyres. The latter current is a westward flow that transports water from the warm pool to the Eastern Pacific Ocean (EPO) and separates the NEC from the SEC (Zhao et al., 2013). A general map showing these currents is given in Fig. A1 in the appendix. The NEC and SEC are close to the ocean surface, and dominated by the trade winds which also blow in the eastward direction. While these currents are well studied, the effect of their inter-annual variability due to the El Niño Southern Oscillation (ENSO) cycle on dFAD drift remains less characterised. In particular, accumulation patterns and dFAD drift are more variable in the western warm pool where the oceanography is complex and influenced more by ENSO phase, than in the EPO (Escalle et al., 2021c; Lopez et al., 2020).

During a positive ENSO phase the tradewinds weaken or even reverse, changing large scale circulation patterns (Horel, 1982) and shifting the warm pool in the zonal direction (Picaut et al., 1996). During an el Niño event, the SEC becomes weaker and the NECC becomes stronger (Wyrtki, 1975). Not only does ENSO affect the large-scale circulation, it also influences mesoscale eddies in the tropical Pacific Ocean. An important mechanism in this study region is the barotropic shear introduced

between the westward SEC and the eastward NECC (Willett et al., 2006). This generates tropical instability waves - westward travelling SST anomalies - and tropical instability vortices - anticyclonic eddies also travelling in the westward direction. This velocity shear weakens during an el Niño event, whereas this shear is strengthened during la Niña years resulting in stronger tropical instability waves and vortices in the latter case (Wang and Fiedler, 2006).

Although considerable research has been done on the el Niño phenomenon and even on its relation to tuna catch (e.g., Lehodey et al., 1997; Lehodey, 2001; Lu et al., 2001; Kumar et al., 2014; OFP-SPC, 2021), the impact of ENSO on the temporal variability of dFAD dispersion in the tropical Pacific Ocean is unknown. The increased attention on the number and distribution of dFADs has raised concerns on the impacts of not only dFADs actively used by fishers, but also those drifting outside of the equatorial fishing zone, which subsequently cause pollution and stranding (Escalle et al., 2019). Spatial management has been suggested as one solution to this problem (Imzilen et al., 2021), but this requires a strong characterisation of ocean flow and its impacts on dFAD dispersion in a region. In parallel, biodegradable dFADs are now being developed (Escalle et al., 2022a) as a possible mitigation against the marine pollution caused by conventional structures, although their degradation is highly variable (Moreno et al., 2023) and likely to be non-linear. Regardless, although all lost dFADs may retain their entire subsurface structure, many are found as simply rafts or even GPS buoys, having presumably broken apart following long drift-times or rough ocean conditions (Escalle et al., 2022b). These remnants experience more surface-driven currents than when they were originally drogued at depth, further complicating spatial management and predictions of high density areas and corridors of loss.

Despite these complications, the drift trajectories of dFADs clearly have the potential to supplement the data provided by drifters for assimilation into ocean circulation models or other applications (Imzilen et al., 2019). It In the Pacific Ocean, it is estimated that 35,000 to 65,000 dFADs are deployed in the Pacific Ocean each year (Escalle et al., 2021b; Lopez et al., 2020). Understanding not only how dFAD drift-profiles may change through their lifetime, but also where and how their drift patterns may differ during their operational lifetime is a critical step towards such an objective. the sustainable management of the fishery. The drift trajectories of dFADs clearly have the potential to be useful in this context, but could also supplement the data provided by drifters for assimilation into ocean circulation models or other applications (Imzilen et al., 2019). However, the access to such real trajectories remains limited in the Pacific Ocean (Lopez et al., 2020; Escalle et al., 2021b). In this context, Lagrangian simulations can be used to explore drift of virtual particles with characteristics similar to dFADs (Escalle et al., 2019; Scutt Phillips et al., 20 .

Here we compare dFAD drift with virtual particle drift solely due to surface currents drift due to two types of virtual particles: i) virtual dFAD drift influenced by currents up to 50 meters, given that dFADs are drogued during their operational lifetime, and thus influenced by deeper currents up to 50 meters ii) virtual particle drift solely due to surface currents. Additionally, we study the correlation between ENSO variability and virtual dFAD displacement, dispersion and metrics that quantify the 'loopiness'

of their trajectories (e.g. Doglioli et al., 2006). This is done by seeding Lagrangian particles in the Pacific Ocean from 15°S to 15°N in a 3D hydrodynamic model using the horizontal flow at multiple depths.

2 Methods

2.1 Data-sets

95 *Ocean Currents*

We used velocity fields from the non-profit organisation Mercator Ocean International (MOi). The horizontal velocity fields at all depths up to 50m were obtained from the MOI_GLO12_WEEKLY_run_for_DAILY_FORECAST product (Mercator Ocean International, 2021). These data are the daily-mean output of a 3D global ocean model assimilated with satellite data and *in-situ* observations, using the GLORYS assimilation scheme. The fields have a horizontal resolution of 1/12° , with 50 100 layers, and are available on an Arakawa C-grid, without interpolation. ~~It is~~ Currently used data spans from 2006-10-11 to 2021-12-31.

This product has been validated globally in Lellouche et al. (2013), where it is found that the hydrodynamic data agrees well with independent observations. A very recent publication (Fritz et al., 2023) also confirms that the product captures the 105 circulation in the Indonesian Throughflow and Western Tropical Pacific well. However, it is known that the tropical SEC is slightly too strong in the MOi data when compared to ~~in-situ 15~~in-situ 15m drogued drifters (Lellouche et al., 2021), which may result in ~~an overestimation of~~ a slight overestimation of virtual dFAD displacement and loopiness for virtual particles released in the SEC. ~~Currently used data spans from 2006-10-11 to 2021-12-31~~There is no dedicated validation of the full Tropical Pacific circulation, but our analysis shows that the Nino3.4 index in the hydrodynamic data agrees well with sea 110 surface temperature observations (see below).

Sea Surface Temperature

SST data is taken from the NOAA Physical Sciences Laboratory. The Niño 3.4 SST Index is used (National Ocean and Atmospheric Administration, 2022), which is derived from SST values from the Met Office's Hadley Centre Sea Ice and Sea 115 Surface Temperature data set (HadISST1). For data past 1982, these values are a mixture of *in-situ* measurements and satellite data (Rayner et al., 2003; National Centre for Atmospheric Research, 2022). The NOAA data-set gives the spatial mean of the SST over the Niño 3.4 region (from 5°S to 5°N and 170° to 120°W (Kug et al., 2009)). These data currently span from 1870 to April 2022 on a one month interval. To transform these data into the Niño 3.4 index the monthly climatology from 1992 to 2021 is removed.

2.2 Virtual Particle Simulations

The simulations are done using the Lagrangian framework Parcels (Delandmeter and Van Sebille, 2019). The simulation lasts over the entire duration of the MOi data-sets. Trajectories are outputted with a daily temporal resolution. Particles are integrated forwards in time with time-steps of 6 hours using only a kernel for two-dimensional advection. This ~~is the~~ means we do not account for the mass and small scale hydrodynamic properties of dFADs in this study. On the scale of the global ocean this assumption holds, as the dFADs have a very low Stokes number and as such do not have their own inertia (Van Sebille et al., 2020). Time-steps of 6 hours are the limit at which most particles do not travel more than 1 grid cell (1/12°) per time step, satisfying the Courant-Friedrichs-Lewy condition.

Every tenth day in the simulation, 17,016 new particles are released from 15°N to 15°S and 140°E to 75°W, with a fixed 0.5° spacing. Particles are deleted after 30 days due to computational limits. Vertical velocities are not taken into account. Two different configurations are used to simulate virtual particle drift: (1) using a depth averaged flow over the upper 50 meters, to account for dFAD appendages; and (2) using surface flow only~~;~~. As not all depth layers in the data are equally wide, each layer's contribution to the depth-averaged velocity is weighted depending on the layer thickness.

2.3 Data Analysis

2.3.1 ~~Variables~~Metrics

Displacement distance, travel distance and distance ratio

We use the displacement distance to quantify the 'straightness' of a trajectory. Displacement distances are calculated using the Haversine distance: the formula for calculating the shortest distance between two points on earth. By applying this equation between every subsequent position in a trajectory and taking the sum of all of these small distances, an approximation of the travel distance is made:

$$d_{30} \approx \sum_{i=1}^{30} x_{i-1,i}, \tag{1}$$

where d_{30} is the travel distance after 30 days, and $x_{i-1,i}$ is the displacement distance between day $i-1$ and day i. This method becomes more inaccurate if the temporal resolution becomes lower. The distance ratio is obtained through dividing the displacement distance by the travel distance:

$$\eta_{30} = \frac{x_{30}}{d_{30}} \approx \frac{x_{30}}{\sum_{i=1}^{30} x_{i-1,i}}, \tag{2}$$

where η_{30} denotes the distance ratio and x_{30} the displacement distance after 30 days. Ratios closer to 1 imply straight trajectories, whereas ratios closer to 0 imply circulation to the same location.

Loopiness

150 Loopiness can be used to quantify eddy-like behaviour. To obtain the loopiness of a trajectory, first the bearing (angle clockwise from the north) between each point in the trajectory is calculated. Thereafter, the smallest difference between those angles is calculated and those differences are summed up, assigning a loopiness value to a trajectory~~-~~:

$$\theta_{30} \approx \sum_{i=1}^{30} \phi_i, \tag{3}$$

where θ_{30} is the after 30 days, and ϕ_i is the angle difference of the trajectory between day i and day $i-1$. Here, outliers occur
155 when ~~an angle difference~~ ϕ_i comes close to plus or minus π, as this minor difference strongly effects whether the loopiness value increases or decreases. The time step here is sufficiently small that this is a very rare occurrence, such that it does not influence the statistical analysis. The definitions of the metrics described in this section are visualised in Fig. 1.
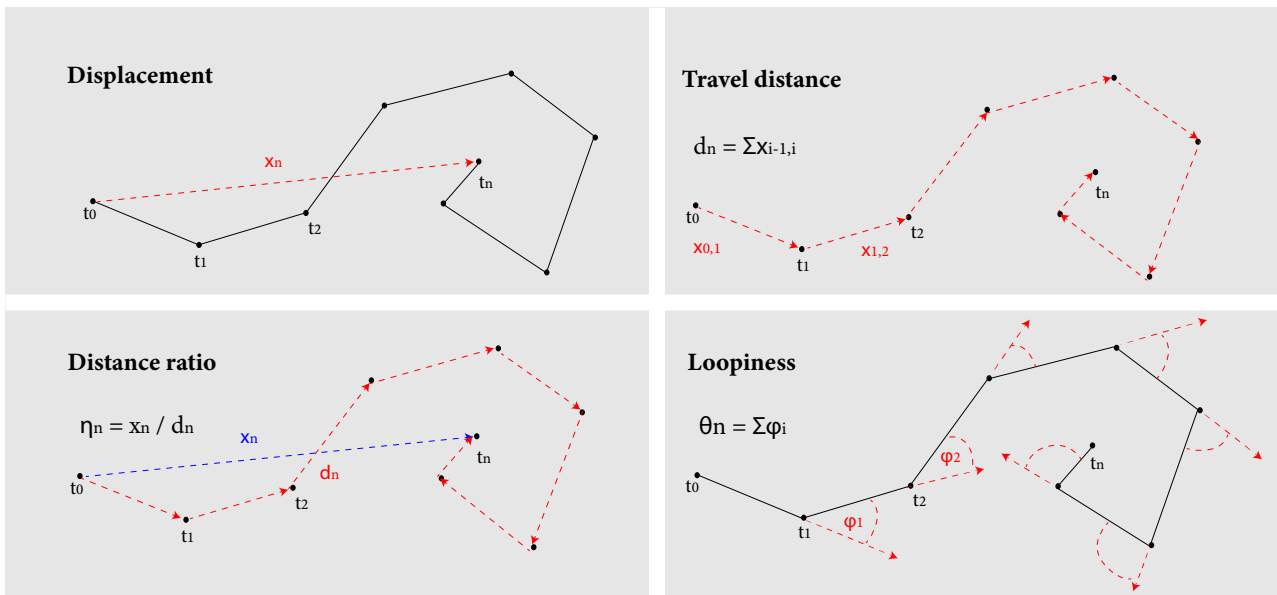

Figure 1. Conceptual visualisation of the metrics used in this study: displacement (top left), travel distance (top right), distance ratio (bottom left) and loopiness (bottom right).

2.3.2 Analysing ENSO variability

To study variability of the virtual dFAD dispersion in relation to ENSO~~composite images of positive and negative ENSO phases~~
160 ~~are created, and variables are correlated to~~, the correlation between the metrics previously described and the Niño3.4 index ~~,~~ is assessed. ~~where the time assigned to a particletrajectory is defined as the time at which a particle is seeded into the domain. Correlations between variables and the Niño3~~First, values from each metric (i.e., virtual particle's displacement, distance ratio,

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and loopiness) are grouped by the month in which the virtual particles were released. Second, a monthly average is calculated for each metric. Then, the correlation between the monthly average and the monthly Nino3.4 index ~~are done with monthly~~
165 ~~data.~~ is assessed using Pearson correlation statistics. Two-sided p-values are calculated under the assumption that the metrics and the Nino3.4 index are drawn from independent normal distributions. Note that every tenth day new particles are released on the same location, hence multiple values for displacement and loopiness are observed each month. A monthly ~~mean~~ average of these values is taken, but due to the irregular time span of months (28-31 days) these monthly averaged values consist of 2,3 or 4 data points.

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~~An ENSO event in this research is defined as a deviation of 1°or higher in the Niño3.4 index that last three months or longer. Within the simulated temporal period there have been two el Niño events satisfying these criteria (2009 and 2015); there have been 3 la Niña events (2008, 2010 and 2020). In the data analysis, these ENSO events are used to create composite figures for the distance ratio and these specifications are not used for the correlation analyses.~~

175 ## 3 Results

3.1 Displacement

Variability of the (depth-averaged) flow field in the tropical Pacific Ocean with respect to ENSO is more dominant in the zonal direction than in the meridional direction, which by extension also holds for dFAD displacement. Fig. 2 shows the correlation (top figure) and linear regression coefficient (bottom figure) between virtual dFAD displacement after 1 month
180 and the ENSO index. Overall, the largest areas with significant correlations are centered around the West and Central Pacific Ocean. The strongest variability is shown at the North Equatorial Countercurrent, with a maximum correlation coefficient of 0.59, from where simulated dFADs can end up 293 ± 30km further away per unit of ENSO change. This is a considerable change, considering ~~the average~~ that a virtual dFAD released from this location travels roughly 1000km in the simulation. This observation of virtual dFADs drifting further east during an el Niño event is in accordance with indications of increased
185 strength of the NECC during el Niño (Zhao et al., 2013). The ~~striped pattern~~ pattern of alternating positive and negative correlations surrounding the NECC ~~could possibly be~~ is likely due to the southward shift in the NECC during the 2009-2010 (Zhao et al., 2013) and 2015-2016 (Zhou et al., 2021) el Niño and the northward shift during la Niña~~, or due to stronger eddies in the NECC during el Niño reducing dFAD displacement. Whether this pattern is actually an ENSO phenomenon is unclear.~~ . However, this latitudinal shift of the NECC does not occur during every ENSO event. Our analysis shows no significant
190 correlations of distance ratio and loopiness in Fig. 3 and 4 in the NECC to explain these patterns with eddy-like behaviour. The South Equatorial Current weakens during a warm event (Wyrtki, 1975), offering an explanation for the strongly negatively correlated blue area, indicating that dFADs originating here are less displaced.

3.2 Mesoscale phenomena

Not all behaviour can be attributed solely to large scale zonal flow, as mesocale phenomena also impact dFAD trajectories. Besides only looking at how far virtual dFADs have displaced, the full trajectory can also be taken into account. In Fig. 3 the ~~ratio between~~ distance ratio between the displacement distance and the travel distance is given, as well as ~~composite images for the el Niño and la Niña events.~~ the correlation and linear regression coefficient to the Niño3.4 index.

~~When comparing la Niña events to averaged conditions, distance ratios tended more towards the extremes in the open ocean.This means that straight trajectories appear even straighter and wavy trajectories meander or spiral even more. This supports the hypothesis that la Niña events mainly amplify the usual flow pattern. In the case of an el Niño, a large difference occurs in the western part of the SEC. Very low ratios are observed in the same region as in Fig. 2 where the displacement was strongly negatively correlated with the ENSO index. This new~~ Fig. 3b shows a large negatively correlated region in the SEC, that indicates lower distance ratios during warm anomalies. This information suggests that in this region, where many Pacific island nations are located, the decrease in virtual dFAD displacement during el Niño events is also due to eddy-like behaviour.
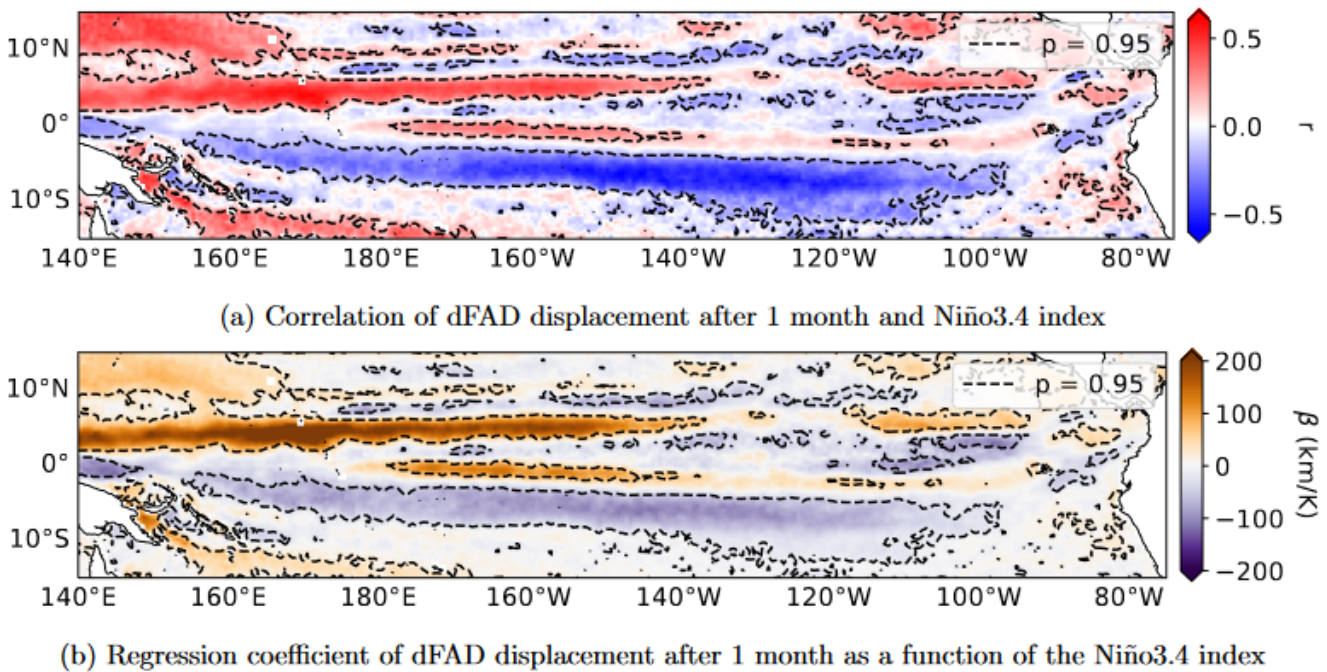

(a) Correlation of dFAD displacement after 1 month and Niño3.4 index

(b) Regression coefficient of dFAD displacement after 1 month as a function of the Niño3.4 index

Figure 2. (a) Correlation between virtual dFAD displacement after 1 month and the monthly Niño3.4 index. Positive correlation are shown in red, whereas blue colours show a negative correlation. The dotted black lines mark regions where correlations are significant (probability values larger than 95%). **(b)** Linear regression coefficient between virtual dFAD displacement after 1 month and the monthly Niño3.4 index. Units are in kmK^{-1}.

The distance ratio in Fig. 3b north and south of the NECCis also very low, which may be attributed to the strong meandering this current experienced in 2010 (Zhao et al., 2013).However, it is important to keep in mind that these composites are comprised
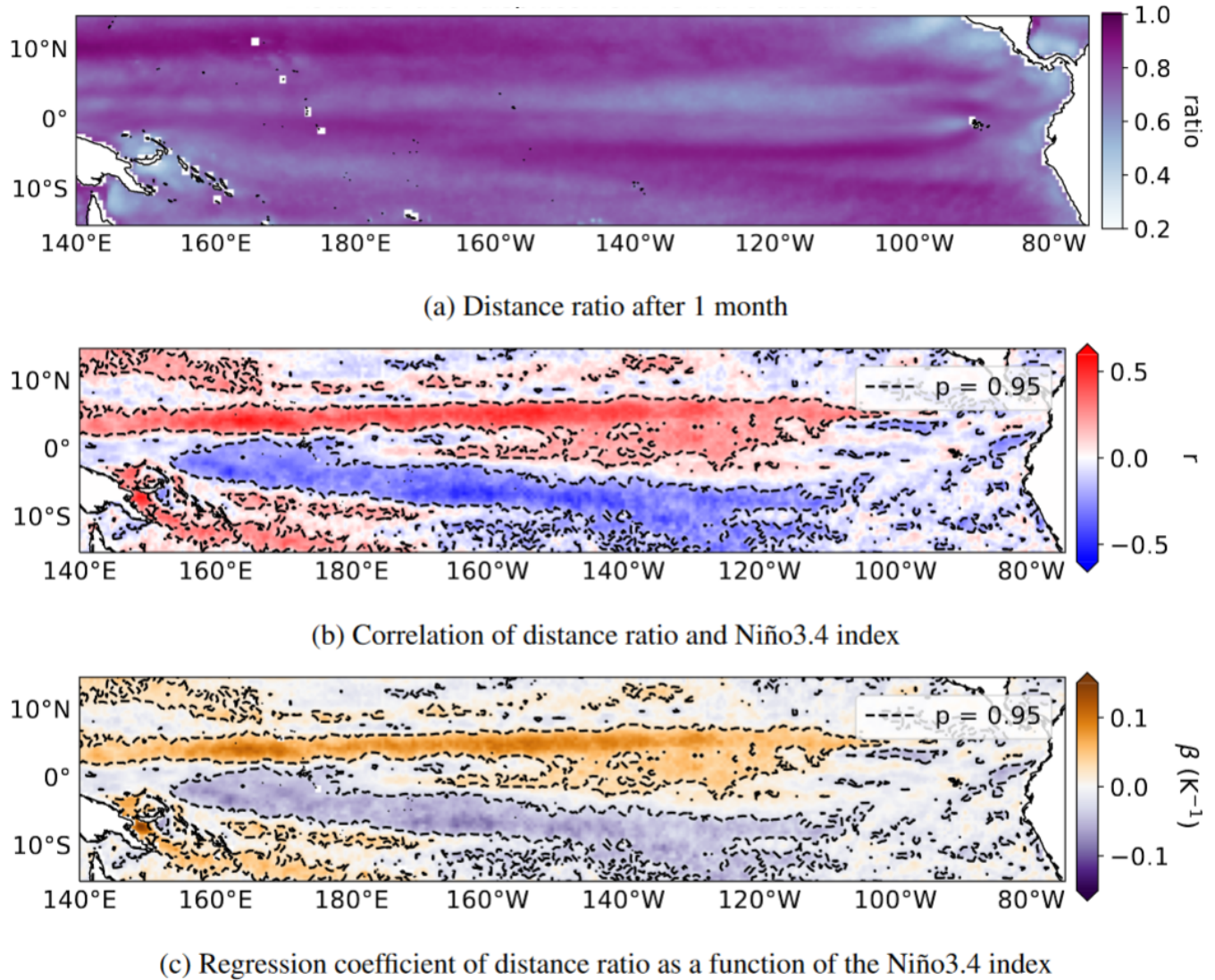

(a) Distance ratio after 1 month

(b) Correlation of distance ratio and Niño3.4 index

(c) Regression coefficient of distance ratio as a function of the Niño3.4 index

Figure 3. Composite figures showing the **(a)** Mean virtual dFAD distance ratio between dFAD displacement and the full trajectory after one 1 monthof travel during **(a)** all 15 years, **(b)** el Niño events and **(c)** la Niña events. Values closer to 1 (in dark purple) indicate nearly straight trajectories, whereas values closer to 0 (in white) indicate stronger circulation. **(b)** Correlation between virtual dFAD distance ratio after 1 month and the monthly Niño3.4 index. Positive correlations are shown in red, whereas blue colours show negative correlations. The dotted black lines mark regions where correlations are significant (probability values larger than 95%). **(c)** Linear regression coefficient between virtual dFAD distance ratio after 1 month and the monthly Niño3.4 index. Units are in K^{-1}, where K is the unit of the ENSO index.

of respectively only two and three events . This means that differences between composite figures do not exclusively showcase ENSO variability, but may also demonstrate noise. A positively correlated region is observed in the NECC, that indicates higher distance ratios during warm anomalies. In the WCPO, this coincides with the region in Fig. 2a, where displacement also shows a strong positive correlation. Together, this shows that virtual dFADs are displaced further and in straighter trajectories during warm events compared to cold events.

Whereas the distance ratio measures how straight particle trajectories are, the loopiness parameter quantifies the total change in direction by taking the sum of all angles during a one month trajectory. The modelled loopiness is shown in Fig. 4a. The most notable feature in this figure is the near anti-symmetry between the Northern and Southern Hemisphere: virtual dFAD loopiness generally is clockwise in the north, but counterclockwise in the south. This can almost certainly be traced back to the direction of the Coriolis acceleration on each Hemisphere. A further feature in this figure are the very low loopiness values near the equator. This may indicate that the virtual dFAD trajectories near the equator go straight, however it is also plausible that this feature is the result of the positive and negative loopiness values corresponding from different langrangian trajectories averaging each other out.

Figs. 4b and 4c show a large area slightly south of the equator where loopiness is significantly negatively correlated to the Niño3.4 index. Judging from the minimum regression coefficientsduring el Niño events, regression coefficients, virtual dFADs seeded in this region can experience more than 20% change in loopiness per unit Kelvin. This region coincides with the region where the distance ratio is close to zero, supporting the observation that during el Niño events, eddy-like behaviour in the southern part of the domain becomes more prevalent. However, a A negatively correlated patch is also observed north-east of the domainin the East Pacific Ocean, where the NECC and the SEC shear. Together with the larger distance ratios observed in the East Pacific Ocean positive correlation of the distance ratio observed in same location in Fig. 3brelative to Fig 3c, this indicates more eddy-like behaviour during la Niña and less during el Niño in this region. These observations seem This observation is in agreement with the theories that tropical instability waves are stronger in this region during a la Niña event (Yu and Liu, 2003; Imada and Kimoto, 2012) and that the associated tropical instability vortices (which grow between 3°N and 8°N in the East Pacific Ocean) are less active during el Niño years (Yu and Liu, 2003).

3.3 Depth-averaged flow versus surface flow

The appendages of a dFAD are meant to slow down the dFAD due to the slower moving currents below the surface. In theory, this opens up possibilities for new interactions that are unseen if only surface currents are considered, which may affect the drift of dFAD rafts after their eventual break down and sinking of their negatively buoyant appendages.

Fig. 5 shows the differences between the two simulations. The displacement figures show mainly the same pattern, though with displacements up to 35% higher over the entire domain when only accounting for surface flow. These higher displacements for surface particles are most likely due to the absence of slower subsurface currents that slow down the virtual dFAD particles.

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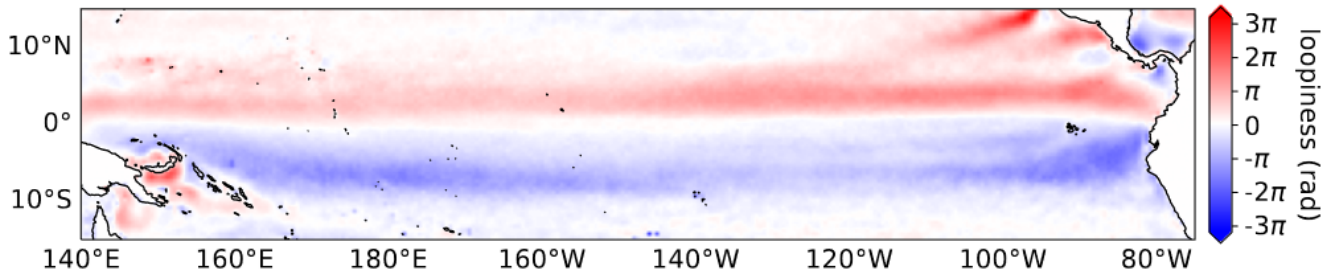
(a) Loopiness after 1 month
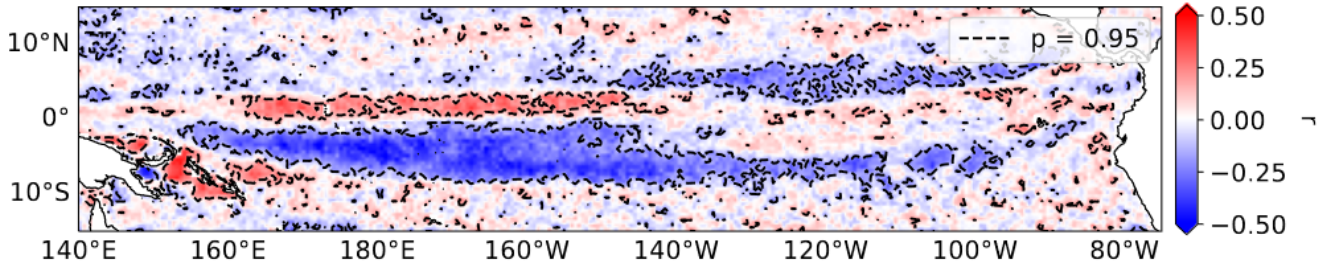
(b) Correlation of Loopiness and Niño3.4 index
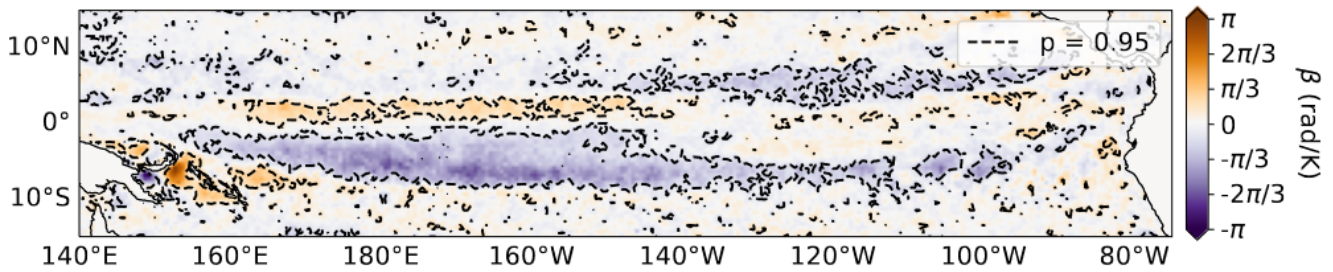
(c) Regression coefficient of Loopiness as a function of the Niño3.4 index

Figure 4. (a) Mean virtual dFAD loopiness in radians after 1 month. Red values mark clockwise motion and blue values mark counterclockwise motion. **(b)** Correlation between virtual dFAD loopiness after 1 month and the monthly Niño3.4 index. Positive correlations are shown in red, whereas blue colours show negative correlations. The dotted black lines mark regions where correlations are significant (probability values larger than 95%). **(c)** Linear regression coefficient between virtual dFAD loopiness after 1 month and the monthly Niño3.4 index. Units are in radK^{-1}, where K is the unit of the ENSO index.

~~Distance ratios during el Niño are nearly identical~~ The correlation of the distance ratio and the Niño3.4 index show similar patterns, in both cases. There is however a notable difference with ~~higher values on the south-east of the domain, where particles appear to travel in generally straighter lines in the simulations with~~ stronger negative correlations that stretch over a larger area in the SEC, indicating that the distance ratio of virtual particles advected by just surface flow ~~. Similar differences are not found~~
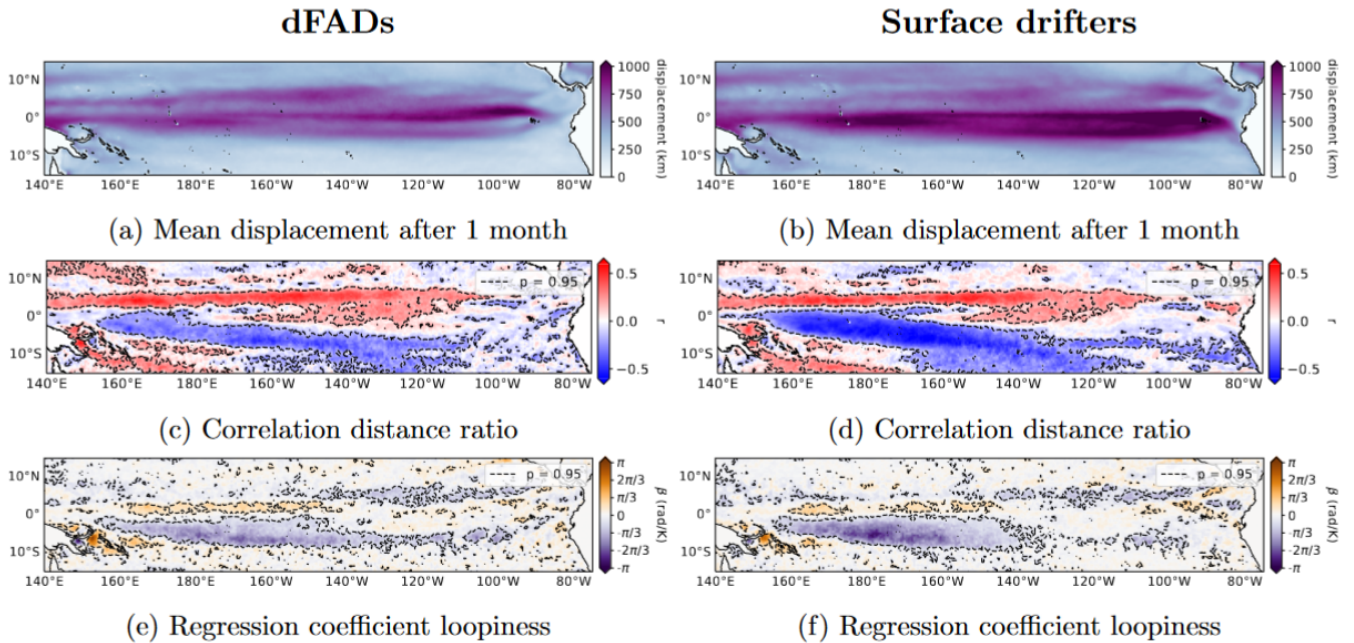
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dFADs   **Surface drifters**

(a) Mean displacement after 1 month   (b) Mean displacement after 1 month

(c) Correlation distance ratio   (d) Correlation distance ratio

(e) Regression coefficient loopiness   (f) Regression coefficient loopiness

Figure 5. Comparison between virtual particle simulations with depth-averaged flow (left) and only surface flow (right). **(a,b)** Mean particle displacement after 1 month of travel. **(c,d)** ~~Distance~~ Correlation of distance ratio ~~between displacement~~ and ~~travel distance after 1 month of travel during el Niño~~the monthly Niño3.4 index. **(e,f)** Linear regression coefficient between ~~dFAD~~ loopiness after 1 month and the monthly Niño3.4 index.

~~in other comparisons.~~ experiences stronger variability with ENSO. Lastly, the regression coefficients of loopiness with respect to the Niño3.4 index are shown in Figs. 5e and 5f. Patterns are similar, with the only difference that the drogued particles in the strongly negatively correlated area experience weaker variability. A likely explanation for this behaviour is that particles with lower velocities travel shorter distances and thus make less loops, which by extension leads to weaker ENSO variability. The behaviour of stronger variability with ENSO of virtual particle drift, advected by only surface currents, shows throughout all metrics used in this research.

4 Discussion and outlook

In this paper, we analysed the drift of virtual dFADs in the tropical Pacific Ocean, using Lagrangian particle tracking. To estimate the effect of their 50m long appendages, we applied two types of simulations. The first was performed using a depth averaged flow over the upper 50m, weighed by layer depth, and the second used only surface flow to advect particles. Using this method, we found that the virtual particles advected by only the surface flow can be displaced up to 35% more than those subjected to the depth-averaged flow and have a stronger loopiness variability with respect to ENSO, but no substantial differ-

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ences between the two simulations were found in patterns across the domain of the tropical Pacific Ocean.

260   Regions identified in this paper as having the highest loopiness and lowest displacement, the south-western part of the WCPO (Papua New Guinea, Solomon Islands, Tuvalu, Kiribati Gilbert Islands), the south-eastern EPO and the northern part of the EPO between 3°N and 8°N, are known areas of dFAD aggregations (Escalle et al., 2021c; Lopez et al., 2020). The zone along the equator, characterised with very high displacements, are known high-speed drift areas of dFADs, where most deployments occurred in the EPO, before dFADs are brought to each side of the equator by divergent currents and they either accumulate in
265   eddies in the northern hemisphere or they are transferred towards to southern WCPO in the southern hemisphere (Lopez et al., 2020).

      In some cases, simulations can be directly compared to observations provided by fisheries on dFAD densities (Escalle et al., 2021a; Scutt Phillips et al., 2019), but these data on dFAD distributions do not cover the entire domain studied in this paper, nor
270   are they publicly accessible. Such data could be used to potentially improve the model, for example improving the calculation of the depth averaged velocity field so that each depth layer is weighed depending on the actual shape of a dFAD instead of only the depth of a layer. Furthermore, scenarios that vary the depth integrated velocity field over individual particle drift-time could be undertaken, adjusting the weight given to each depth layer over time as to simulate the biodegradable aspect of dFADs (appendages becoming shorter over time).

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      Correlations between virtual dFAD displacement and the Niño3.4 index were very high in some locations, up to a maximum of 0.59 in the NECC, indicating strong ENSO variability. Overall, the virtual dFADs tended to be displaced more during an el Niño if seeded in the NECC and less if seeded in part of the SEC below the equator, which can be attributed to the respective strengthening and weakening of the currents during the positive ENSO phase. These two areas with higher correlation between
280   virtual dFAD displacement and ENSO, correspond to areas with lower or higher dFAD aggregation during el Niño, relatively.

      dFAD fishing in the WCPO NECC is generally low, and it decreases even further during el Niño, with the whole fishery moving East (Williams and Ruaia, 2020). The WCPO SEC, indicates negative correlation between virtual dFAD displacement and ENSO, corresponding to an area with higher dFAD fishing effort during el Niño (Williams and Ruaia, 2020). The mech-
285   anisms behind the observed response of the WCPO tuna fishery to el Niño (Williams and Ruaia, 2020; Lehodey et al., 1997; OFP-SPC, 2021) may be related to these same drivers of change in dFAD drift as a result of ocean flow. The high FAD density areas in the southern WCPO (Escalle et al., 2021c) might extend or move east during an el Niño event, although this could not be verified due to the lack of availability of observed trajectories covering this region.

      Different patterns in displacement, eddy distance ratio, loopiness and pathways often follow zonal lines, as the dominant
290   currents in the open ocean of this domain travel mainly in the zonal direction. Dispersion in the zonal direction may however be overestimated in this research as the tropical SEC is too strong in the MOi data when compared to *in-situ* 15m drogued

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drifters (Lellouche et al., 2021).

In the EPO, stronger loopiness ~~was observed during a la Niña event~~ and lower distance ratios between the NECC and SEC ~~. This loopiness mostly disappeared during el Niño events. A possible explanation could be that during~~ (north of the equator) were observed during colder conditions. This coincides with the region where during la Niña, stronger generation of tropical instability vortices takes place, due to ~~to~~ higher velocity shear between the eastwards and westward currents (Yu and Liu, 2003; Imada and Kimoto, 2012). In the WCPO, a region of stronger eddy-like behaviour was found in the SEC, that became stronger during ~~el Niño due to~~ warm conditions, with decreased distance ratios and increased loopiness. South of the equator, weak TIW patterns exist (Xue et al., 2020), but they should have an opposite effect on our eddy statistics, compared to our observations, which suggests TIWs are not the driving mechanism behind these patterns in the WCPO SEC. The mechanisms behind this behaviour ~~, either of physical nature or artifacts from the strong SEC in the MOi data,~~ require further research, as most research on eddy polarity statistics in this region do not focus on the tropics. The main findings and reasonings of this research are summarised in Fig. 6.

In determining the variability of virtual dFAD dispersion with respect to ENSO, correlations were significant, but only a few ENSO events ~~were taken into account~~ took place in the time frame of this research. The results consist of data with ~~two~~ three el Niño and three la Niña events ~~(one of which only lasted three months above 1°)~~. As the currently used data sets are still being expanded upon for the foreseeable future, more events may eventually be captured and analysed leading to even more robust results.

Moreover, the phase of the ENSO event is mostly ignored throughout this paper. In some cases, ENSO related variability of Pacific Ocean currents distinguishes between the developing stages and mature stages of an el Niño/la Niña event (e.g. Wu et al., 2016; Zhai and Hu, 2012; Zhao et al., 2013; Tan and Zhou, 2018) and the delay of its effects depending on the location. In this paper, where most calculated ~~variables~~ metrics in time series data are already averaged or summed over the time of a month, there is a 1 month phase lag within the calculation of one value assigned to a trajectory. Hence, it is arbitrarily chosen as the time of virtual dFAD deployment. The effects of phase lag of calculated values such as displacement, loopiness and distance ratio with respect to the Niño3.4 index have not been investigated here.

Only a fraction of the amount of data MOi has to offer has been used in this study, whereas their products have a lot more to offer for further research regarding dFAD drift. Future research could also focus on the actual link to tuna catch or effectiveness of certain release locations. Moreover, it could be interesting to track bio-activity along the dFAD trajectory, such as chlorophyll-a concentrations, to see if and how dFADs, released in particular areas, travel through nutrient-rich areas. Furthermore, we used the distance ratio and loopiness to quantify whether dFADs appear to be travelling in an eddy, classifying

it as 'eddy-like behaviour'. Such data can be further built upon by tracking the along trajectory eddy diffusivity, in order to determine the mechanisms behind specific types of trajectories.
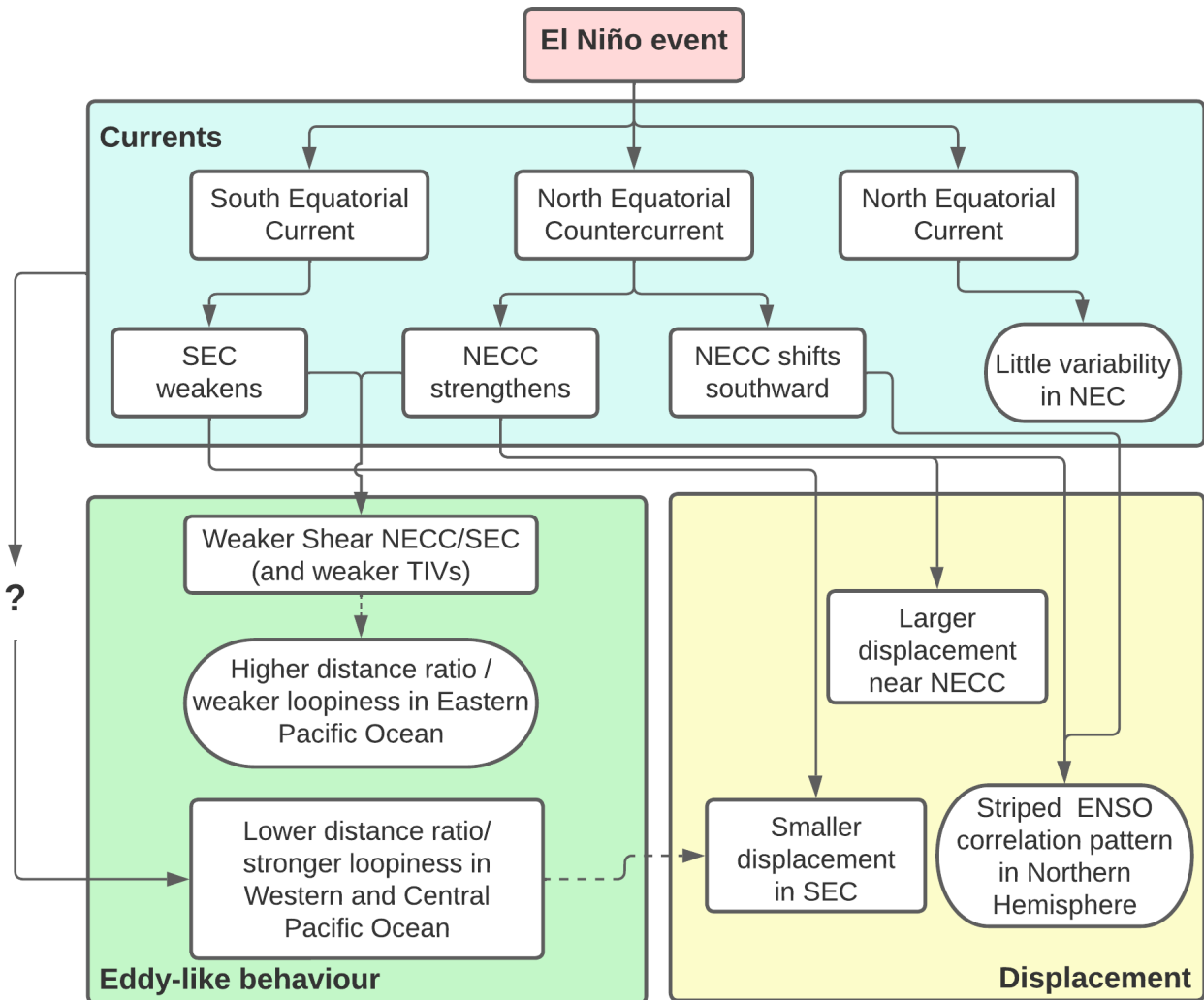

Figure 6. Diagram showing the general findings and considerations of this paper. Arrows mark how findings and locations influence each other. Dashed arrows denote possible connections.
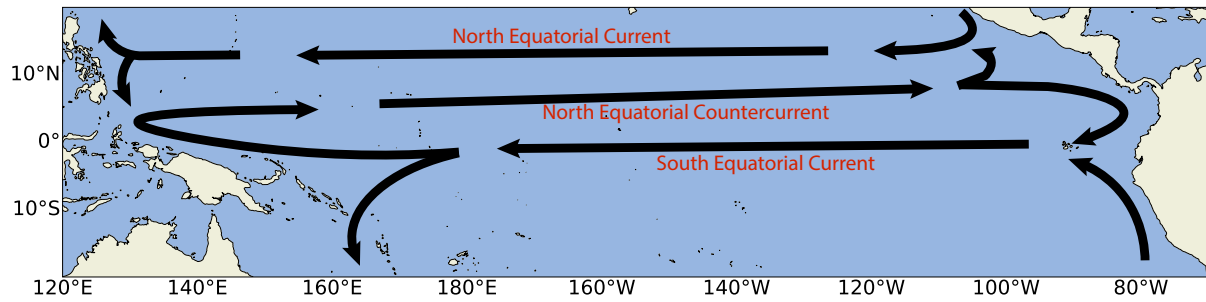
Appendix A

Figure A1. General overview of currents in the tropical Pacific Ocean. The three main currents in the open ocean are the North Equatorial Current, the North Equatorial Countercurrent and the South Equatorial Current.

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