

TECHNICAL AND COMPLIANCE COMMITTEE

Nineteenth Regular Session 20 – 26 September 2023

Pohnpei, Federated States of Micronesia

Ecosystem and Climate Indicator Report Card from SC19

WCPFC-TCC19-2023-IP12¹ 30 July 2023

SPC-OFP

¹ This paper is Annex 2 to a paper that was posted to SC19 as SC19-2023-EB-WP01 *Ecosystem and Climate Indicators*

Annex 2 Ecosystem and Climate Indicator Report Card

Details on the calculations for each indicator in Report Cards 1 to 3 are provided below. Code, data, and associated figures and results for each indicator are available in the GitHub repository for the paper: github.com/PacificCommunity/OFP-FEMA-ecosystem-indicators.

Report Card 1. Environment Indicators

All environmental indicators were calculated from outputs of the Bluelink Ocean ReANalaysis 2020 (Chamberlain et al. 2021), a three-dimensional, physical ocean model with a spatial resolution of 1/12°. Monthly outputs were used to allow averaging over seasons, when required by an indicator. The code used to generate indicators from pre-processed netcdf output files from BRAN2020 can be found at the GitHub repository for this paper (see link above).

A.1 Sea Surface Temperature Anomalies

Sea surface temperature (SST) anomaly was calculated across three spatial extents. In all three cases, the annual value was the mean anomaly of all cells within the spatial extent, from a baseline mean across the period 1993-2021. For the WCPO SST anomaly, this spatial extent was bounded by a square with corners at 50°N, 130°E and 50°S, 150°W (see Figure 1 in SPC-OFP 2021). The WCPO equatorial SST anomaly included only cells bounded by the box with corners at 5°N, 130°E and 5°S, 150°W. In the case of the warm pool extent SST anomaly, the spatial extent of cell anomalies changed each year. Following a typical characterisation of the warm-pool extent, only those cells that exceed a mean sea surface temperature of 29°C during the period November to April were included in anomaly of cells included in this extent, from their respective 1993-2021 baseline, was then calculated annually for the period November to April.

A.2 Warm Pool Indices

Each year, the extent of the warm pool was calculated using the method described above. In the case of the mean warm pool size, the number of cells with a mean sea surface temperature greater than 29°C during November to April was used to provide the approximate area encompassed by the warm pool each year. The eastern boundary of the warm pool was calculated following a similar methodology to Qu and Yu (2014) and others, where strong changes in sea surface salinity (SSS) across the equator were used to indicate the presence of a barrier layer between increased fresh water in the warm pool meeting colder, high salinity water from the east. Mean SSS between 2°S and 2°N was calculated during the November to April period, and the centre of the largest longitudinal change across a 10° window identified as the eastern limit of the warm pool. The mean warm pool mixed layer depth (the depth at which water mixing results in uniform buoyancy of a particular value) was simply taken directly from BRAN2020, and averaged over the extent of the warm pool during the period November to April each year.

A.3 Climate Indices

Here, we have presented two climate indices which relate to changes in the WCPO ecosystem. The Oceanic Niño Index (ONI) tracks three-month averaged SST anomalies across regions of the equatorial Pacific from a moving 30-year average temperature, and one method of identifying likely El Niño or La Niña events. The Interdecadal Pacific Oscillation index (IPO) measures longer-term climate cycles affecting the extent of the Pacific basin, and switches phases roughly each 15-30 years. Positive phases are associated with increased warming in the tropics and cooler northern Pacific climate, and negative phases are associated with cooler temperatures in the tropics and increased temperatures in the higher latitudes.

Report Card 2. Annual Tuna Catch & Fishing Effort indicators

A.4 Annual Tuna Catch

These indicators describe trends in annual catch estimates (in metric tonnes) of the four main tuna species (skipjack, yellowfin, bigeye and albacore) targeted within the WCPFC Convention Area (WCPFC-CA), between 1990 and 2022, inclusive. Data for the calculations were extracted from SPC's 'a_model' database, a collation of S_BEST, L_BEST, and P_BEST catch data aggregated at 5° x 5° resolution for all fishing gears, and S_BEST and L_BEST containing aggregated, raised catch data from the purse-seine fishery at 1° x 1°, and the longline fishery at 5° x 5°, respectively. See Hare et al. (2023) [SC19/SA-WP-06] for a compilation of all fishery indicators for these target tunas.

A.5 Fishing Effort

Data to characterize trends in fishing effort were extracted from SPC's S_BEST and L_BEST databases from 1990-2033, inclusive, for purse seine (PS) and longline (LL) catch and effort data, respectively. These databases contain aggregated, raised fishing effort across the WCPFC-CA. We focused on purse seine and longline data as they represent the major gear sectors for the region. For the purse seine fishery, the individual fishing set was considered the metric of effort, while for longline, effort was defined as the number of hooks fished.

The central tendency of purse seine fishing effort was defined here by the 'centre of gravity', i.e. the mean location (latitude and longitude) of fishing effort. This was calculated by year for each fishing mode i.e. 'unassociated' free-school sets (UNA) versus 'associated' sets (ASS). We present only the annual longitudinal centre of gravity for purse seine, as the fishery remains relatively stable latitudinally year on year. It should be noted that for this analysis, associated sets refers to sets made on drifting FADs and drifting logs or debris; this does not include sets made around whales or whale sharks, nor does it include anchored FAD sets.

The central tendency indicators were not calculated for the longline fishery because of the diversity in targeted species and the areas associated with different targeting behaviours. At this time, a measure of central tendency for the longline fishery was not expected to be an informative indicator of ecosystem dynamics.

In addition to the central tendency of fishing effort, area occupied by the purse seine and longline fisheries was calculated. Area occupied is a measure of the distribution of effort across the spatial domain of the WCPFC and was calculated as the sum of the area (in km²) of unique 1° x 1° cells fished by the purse seine fishery and 5° x 5° cells fished by the longline fishery, in each year evaluated.

With growing interest in tracking changes in the distribution of purse seine effort inside and outside EEZs within the WCPFC-CA, we include a new effort indicator this year, representing the proportion of purse seine sets made in High Seas areas, disaggregated by fishing mode. High Seas areas included in the calculations comprise the I1, I2, I3, I4, I5, I6, I7, I8, I9, H4, H5 regions (Figure A1).

Figure A1. WCPFC-CA High Seas regions and boundaries.



Report Card 3. Biology & Bycatch Indicators

A.6 Tuna Condition

The mean fork length (cm) of skipjack tuna was calculated annually from all length measurements recorded for longline, purse seine and pole-and-line catches made in the WCPFC-CA between 1990 and 2022, inclusive. Length data were drawn from observer and port sampling records, in this case contained in SPC's 'BioDaSys', 'OBSV_MASTER' and 'Tufman2' databases. Following the methods used for the fishing effort indicators (see section A.5) we focussed our attention on the purse seine and longline data as they represent the major fisheries in terms of catch, and were available across the full 33-year time series. Where required, published 'conversion factors' were used to convert length measurements to fork length (UF) in cm. These conversion factor equations are updated as new data comes to hand, and are housed in an online database managed by SPC. We refer readers to Macdonald et al. 2023 [SC19/ST-IP-04] for an update on progress on this conversion factor work.

The mean fork length (cm) of yellowfin and bigeye tuna caught in the longline fishery was calculated annually from all length measurements recorded for each species within the WCPFC-CA between 1990 and 2022, inclusive. The length data were again drawn from observer and port sampling records contained in SPC's 'BioDaSys', 'OBSV_MASTER', 'FISH_MASTER' and 'Tufman2' databases. We focussed on the longline data for yellowfin and bigeye, as this gear typically selects for larger individuals than purse seine, placing a lower bound on the length range considered. This allowed us to maximise precision, while minimising potential gear-related bias in tracking shifts in mean length through time. As for skipjack, length measurements were converted to fork length (UF) in cm where required using published conversion factors for each species.

Mean fish condition, defined by the average relative condition factor $K_{rel} = WW/aUF^b$ (where WW is an individual's whole weight (kg) and aUF^b is the model predicted whole weight at fork length UF (cm)) was calculated annually for skipjack, yellowfin and bigeye tuna separately, based on length and weight data from longline catches made across the WCPFC-CA between 1990 and 2022, inclusive. The data were drawn from observer and port sampling records contained in SPC's 'BioDaSys', 'OBSV_MASTER', 'FISH_MASTER' and 'Tufman2' databases.

Published conversion factors were again used to convert length measurements to fork length (UF) in cm, and weight measurements to whole weight (WW) in kg.

For each species, we elected to model predicted weight from the longline records only. This decision was based around two points. i) Data coverage: the broad spatial and temporal extent of coupled length and weight measurements available from the longline fishery provide the most reliable estimates for calculating K_{rel} . ii) Mismatch in scales: given the different size selectivities, areas fished and length of time series available for longline, purse seine and pole-and-line gears, there is potential for the shape of the length-weight curve to differ among gears/areas/time periods fished. Therefore, by fitting our models to the longline data only we aimed to reduce these possible biases in monitoring changes in fish condition across the 1990 to 2022 time series. We note that new sampling initiatives are being developed to enhance data collection on purse seine vessels, and as further data becomes available, gear-to-gear comparisons could be reported in future iterations of these Report Cards.

Fat content represents the percentage of lipids in the tuna flesh, and we consider this a potentially useful second indicator of tuna condition that complements the measurement of K_{rel} . The percentage of fat is measured using the Distell's fish 'fatmeter' model 692 by a simple contact of the instrument's sensor on the skin of the fish. Collection of fat content data on tropical tunas is now part of routine biological sampling tasks during PTTP tuna tagging cruises, and the dataset is growing steadily. Fat content is dependent on fish size; hence to avoid introducing bias, only skipjack, yellowfin and bigeye tuna measuring 40-60cm fork length were used to calculate annual mean fat content by species.

A.7 Bycatch Species

The observer and aggregate effort datasets used to estimate the amount of catch for the bycatch species were extracted from SPC data holdings. The overall approach was to estimate stratified catch rates using a combination of presence/absence models and bootstrap sampling for catch when present, and then to use these catch rates to estimate bycatch for unobserved sets. Recorded

catches were used directly for observed sets, and assumed to be known without error.

For purse seine, the methods are fully described in Peatman and Nicol (2021), and a summary of the approach is provided here. The estimates cover the large-scale equatorial purse seine fishery operating in the WCPFC-CA. Bycatch estimates were not generated for purse seine fleets for which SPC holds limited representative observer data, namely small-scale domestic fisheries of Indonesia, Vietnam and the Philippines, and purse seiners operating in temperate waters. Bycatch estimates were generated in units of individuals for billfish, sharks and rays, with finfish bycatch estimated in units of metric tonnes. These units match those most commonly used by observers when recording catch volumes of the respective species groups and were considered to provide the most accurate dataset of observed catches in SPC's purse seine observer data holdings.

Presence/absence models were fitted to observer data using Generalised Estimating Equations (GEEs) with year, sea-surface temperature (SST – Reynolds et al. 2002), and categorical variables for quarter and school association as explanatory variables. The fitted presence/absence models were used to estimate the probability of presence for a given estimation group and strata (combinations of year, quarter and school association). The volume of catch when present was estimated by bootstrap sampling from sets with observed captures, stratified by association type. Estimates of the overall bycatch rate were then obtained for each estimation group and strata by taking the product of the probability of presence and the volume of catch when present. As such, the units of bycatch rate were numbers or metric tonnes per set. The estimated catch rates were then applied to the number of unobserved sets in each strata, to calculate unobserved bycatch. The estimates of unobserved bycatch were then combined with recorded bycatch from observed sets to give estimates of total bycatch.

For longline, the methods are fully described in Peatman and Nicol (2020), and a summary of the approach is provided here. The estimates cover longline fishing from 2003 to 2018 in the WCPFC-CA, including the region overlapping the IATTC Convention Area. Catch estimates do not include catches from the domestic longline fisheries of the Philippines, Vietnam and Indonesia, referred to in this report as 'west-tropical domestic fisheries', as SPC holds little representative observer data for these fisheries. Catch estimates also do not include former shark-targeted longline fisheries in the Papua New Guinea (PNG) and Solomon Islands (SB) EEZs as these fisheries are not included in aggregate longline catch and effort data held by SPC.

Hooks between float (HBF) specific aggregate catch and effort data, i.e. 'L_BEST_HBF' data, were used to estimate the proportions of aggregate effort data by HBF categories. K-means clustering was applied to aggregate longline catch data to partition longline effort into groups with similar species compositions.

GEEs were again used to model catch rates with year, sea-surface temperature (SST), HBF, and categorical variables for flag, and the species composition cluster for the 'L_BEST' strata as explanatory variables. A simulation modelling framework was used to estimate catches. First, the effort dataset for catch estimation was generated by aggregating HBF-specific effort surfaces to a resolution of year, SST, HBF, catch composition cluster, flag and region. Then estimated catches were obtained by taking the product of the catch rates and the effort.

References

Chamberlain, M. A., Oke, P. R., Fiedler, R. A. S., Beggs, H. M., Brassington, G. B., and Divakaran, P. in review. Next generation of Bluelink ocean reanalysis with multiscale data assimilation: BRAN2020, Earth Syst. Sci. Data Discuss. [preprint], https://doi.org/10.5194/essd-2021-194.

Hare, S., Pilling, G. and Williams, P. 2023. A compendium of fisheries indicators for target tuna stocks in the WCPFC Convention Area. 19th Meeting of the Scientific Committee of the Western and Central Pacific Fisheries Commission, WCPFC-SC19, Koror, Palau, 16-24 August 2023, WCPFC-SC19/SA-WP-06. https://meetings.wcpfc.int/node/19354.

Højsgaard, S., Halekoh, U. and Yan, J. 2005. The R package geepack for generalized estimating equations. Journal of Statistical Software 15(2):11.

Hu, S. and Fedorov, A.V. 2019. The extreme El Niño of 2015–2016: The role of westerly and easterly wind bursts, and preconditioning by the failed 2014 event. Climate Dynamics 52(12): 7339–7357.

Macdonald, J., Williams, P., Hoyle, S., Chang, S.K., Roupsard, F., Ghergariu, M., Nguyen Cuu, S., Hosken, M., Sanchez, C., Contreras, R.B.B. and Nicol, S. 2023. Project 90 update: Better data on fish weights and lengths for scientific analyses. 19th Meeting of the Scientific Committee of the Western and Central Pacific Fisheries Commission, WCPFC-SC19, Koror, Palau, 16-24 August 2023, WCPFC-SC19/ST-IP-04. https://meetings.wcpfc.int/node/19346.

SPC-OFP 2021. WCPO ecosystem and climate indicators from 2000 to 2020. 17th Meeting of the Scientific Committee of the Western and Central Pacific Fisheries Commission, WCPFC-SC17, Online meeting, 11-19 August 2021. WCPFC-SC17-2021/EB-IP-09. https://meetings.wcpfc.int/node/12606.

Peatman T. and Nicol S. 2020. Updated longline bycatch estimates in the WCPO. 16th Meeting of the Scientific Committee of the Western and Central Pacific Fisheries Commission, WCPFC-SC16, Online meeting, 12-20 August 2020, WCPFC-SC16/ST-IP-11. https://meetings.wcpfc.int/node/11690.

Peatman T. and Nicol S. 2021. Updated purse seine bycatch estimates in the WCPO. 17th Meeting of the Scientific Committee of the Western and Central Pacific Fisheries Commission, WCPFC-SC17, Online meeting, 11-19 August 2021, WCPFC-SC17/ST-IP-06. https://meetings.wcpfc.int/node/12542.

Qu, T. and Yu, J.Y. 2014. ENSO indices from sea surface salinity observed by Aquarius and Argo. Journal of Oceanography 70(4): 367–375.

Reynolds, R.W., Rayner, N.A., Smith, T.M., Stokes, D.C. and Wang, W. 2002. An improved in situ and satellite SST analysis for climate. Journal of Climate 15: 1609–1625.

Roxy, M.K., Dasgupta, P., McPhaden, M.J., Suematsu, T., Zhang, C. and Kim, D. 2019. Twofold expansion of the Indo-Pacific warm pool warps the MJO life cycle. Nature 575: 647–651.

Report Card 1. Environment Indicators

Indicator	Description ?	Notes	Time-series
Sea Surface Temperature Anomalies (ANNEX 1 - A.1) Mean/Reference Value — — Central 50% of data range			
Annual SST Anomaly	Mean annual SST anomaly (°C) across WCPO area	 Derived from ocean models WCPO area western limit of 130°E Anomaly from mean temperature 1993-2021 	1990 1995 2000 2005 2010 2015 2020 0.09 -0.09
	Mean annual SST anomaly (°C) across WCPO equatorial zone	 Derived from ocean models Equatorial zone 5°S-5°N Anomaly from mean temperature 1993-2021 	0.24
Nov-Apr Warm-pool SST Anomaly	Mean annual SST anomaly (°C) within warm-pool extent	 Derived from ocean models Warm-pool defined by mean Nov- Apr temperature > 29°C 	0.24
Warm-pool Indices (ANNEX 1 - A.2)			
Mean Size of Warm- pool	Approximate size of warm-pool in millions of km ²	 Derived from ocean models Warm-pool defined by mean Nov- Apr temperature >29°C 	
Eastern Limit of Warm-pool Boundary	Longitude of strongest sea surface salinity boundary	 Derived from ocean models Boundary defined as largest change over 10° distance 	175E 157E
Mean Warm-pool Mixed Layer Depth	Mean depth (m) of the mixed layer within warm-pool	 Derived from ocean models Layer over which water temperature is homogenous 	43.8 42.9 41.9
Climate Indices (ANNEX 1 - A.3)			
Oceanic Niño (ONI) and Interdecadal Pacific Oscillation (IPO) Index	ONI indicates SST anomalies in the Niño 3.4 region during Nov-Jan each year IPO represents long-term oscillation between El Niño favourable and La	 ONI values > 0.5 indicative of El Niño events, values < -0.5 indicative of La Niña IPO values > 0 indicative of more El Niño events, < 0 indicative of more La Niña events 	0.5 0.5 0.5 1990 1995 2000 2005 2010 2015 2020
l		 Time series from 1993-2021 	

Report Card 2. Annual Tuna Catch & Fishing Effort Indicators



Report Card 3. Biology & Bycatch Indicators

