

Preliminary 2024 stock assessment of yellowfin tuna in the Indian Ocean

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Executive Summary

This report presents a preliminary stock assessment for Indian Ocean yellowfin tuna (*Thunnus albacares*) using Stock Synthesis 3 (SS3). The assessment uses an age-structured and spatially-explicit population model and is fitted to catch rate indices, length-composition data, and tagging data. The assessment covers 1950 – 2023 and represents an update of the previous assessment model, taking into account progress and improvements made since the previous assessment. The assessment assumes that the Indian Ocean yellowfin tuna constitute a single spawning stock, modelled as spatially disaggregated four regions, with 21 fisheries. Standardized CPUE series from the main longline fleets 1975 – 2020 were included in the models as the relative abundance index of exploitable biomass in each region. The CPUE indices from EU Purse seine sets on free schools were included in a subset of models with the spatial and fleet structure revised to better accommodate the distribution and size structure of the purse seine fisheries. Indices based on associative and non-associative dynamics of yellowfin tuna with floating objects were also available, and the utility of these indices was examined in the assessment. Tag release and recovery data from the RTTP-IO program were included in the model to inform abundance, movement, and mortality rates.

1 Introduction

Prior to 2008, Indian Ocean (IO) yellowfin tuna (*Thunnus albacares*) was assessed using methods such as Virtual Population Analysis (VPA) and production models (Nishida and Shono 2007, 2005). In 2008, a preliminary stock assessment of IO yellowfin tuna was conducted using MULTIFAN-CL (Langley et al. 2008) enabling the integration of the tag release/recovery data collected from the large-scale tagging programme conducted in the IO in the preceding years. The MULTIFAN-CL assessment was revised and updated in the following years (Langley et al. 2009; Langley, Herrera, and Million 2010, 2011, 2012).

In 2015, the assessment of IO yellowfin tuna was implemented (Langley 2015), which used the Stock Synthesis 3 modelling platform (Methot and Wetzel 2013). SS3 is conceptually very similar to MFCL and the two platforms have yielded similar results. On basis of that assessment, the yellowfin tuna stock was determined to be overfished and subject to overfishing. At its 20th meeting, the Indian Ocean Tuna Commission (IOTC) adopted an Interim Plan for Rebuilding the Indian Ocean Yellowfin Tuna Stock (Res. 16/01).

The SS3 assessment was updated in 2016 (Langley 2016) and was revised and updated in 2018 (Fu et al. 2018). These assessments utilised new composite longline CPUE indices derived from the main distant water longline fleets, replacing the Japanese longline CPUE indices used previously. The 2018 assessment also included a comprehensive analysis of the main assumptions of the stock assessment. A model ensemble covering major components of structural uncertainty was used to characterise the stock status. The assessment estimated that the spawning stock biomass in 2017 was below SSB_{MSY} , and that fishing mortality was above F_{MSY} . Therefore, the stock status was determined to remain overfished and experiencing overfishing.

An external review of the 2018 assessment provided recommendations to improve model parametrisations (Methot 2019). An attempt was made to update the assessment in 2019, with extensive investigations of alternative spatial structures, data weighting and biological parameters (Urtizberea et al. 2019). Further analysis was conducted in 2020 to refine the process of model selection through an objective scoring system based on diagnostic metrics (Urtizberea et al. 2020).

The most recent assessment was conducted in 2021 (Fu et al. 2021), which also used SS3 as the modelling platform and was based on the four area spatial configuration as in 2018. This recent assessment included an standardised CPUE series from the main longline fleets as the main index, but also tested the inclusion of EU purse seine indices, operating on free schools and floating objects, and an index from the Maldivian pole and line fishery. A range of exploratory models were presented to address issues in observational datasets, improve the stability of the assessment model, and explore the effects of alternative model assumptions. Overall stock status estimates do not differ substantially from the 2018 assessment, estimating $SSB/SSB_{MSY} = 0.78$ and $F/F_{MSY} = 1.27$ for the last model year, which suggest that the stock is overfished and experiencing overfishing.

An external review of the 2021 assessment was carried out in 2023 (Maunder et al. 2023), and provided a set of recommendations for the next assessment implementation. This report documents the next iteration of the stock assessment of the IO yellowfin tuna stock for consideration at the 26th WPTT meeting. This stock assessment includes fishery and biological data up to the end of 2023 and its configuration is based on the 2021 assessment. It implements an age- and spatially-structured population model using SS3 (v3.30.22.1) and incorporates some revisions made by the last external review.

2 Background

2.1 Biology

Yellowfin tuna is a cosmopolitan species distributed mainly in the tropical and subtropical oceanic waters of the three major oceans, where it forms large schools. Spawning occurs mainly from December to March in lower latitudes with warmer waters and mesoscale oceanographic activity ([Muhling et al. 2017](#)), with the main spawning grounds west of 75°E. However, spawning activity has also been reported in the Oman Sea ([Hosseini and Kaymaram 2016](#)), Bay of Bengal ([Kumar and Ghosh 2022](#)), off Sri Lanka and the Mozambique Channel, and in the eastern IO off Australia Nootmorn, Yakoh, and Kawises ([2005](#)). The size at 50% maturity for this species in the IO was initially estimated at around 75 cm based on cortical alveolar stage ([Zudaire et al. 2013](#)), but an updated study suggests that it might be at a larger size (~ 101 cm) ([Zudaire et al. 2022](#)). Tag recoveries provide evidence of large movements of yellowfin tuna within the western equatorial region; however, few observations of large-scale transverse movements in the IO have been reported ([Gaertner and Hallier 2015](#)). Yellowfin dwell preferentially in the surface mixed layer and the thermocline ([Pecoraro et al. 2017](#)), above 200 m approximately ([Sabarros, Romanov, and Bach 2015](#)).

This species has a high metabolic rate and, therefore, it requires large energy supplies to fulfill the bioenergetics demands for movement, growth, and reproduction ([Artetxe-Arrate et al. 2021](#)). Feeding behaviour is largely opportunistic, with a variety of prey species being consumed, including large concentrations of crustacean that have occurred recently in the tropical areas and small mesopelagic fishes ([Krishnan et al. 2024](#); [Roger 1994](#); [Duffy et al. 2017](#)). Recent growth studies have generally supported a two-stanza growth curve, with a slow initial growth phase up to ~ 60 cm followed by much faster growth ([J. H. Farley et al. 2023](#)). In addition, differences in mean length-at-age have been identified between males and females for fish older than four years. Environmental variability in the IO impacts the abundance and catch rates of this species. A significant negative association between the Indian Ocean Dipoles (IODs) and the catch rates of yellowfin tuna with a periodicity of approximately four years was observed ([Lan, Evans, and Lee 2013](#); [Lan, Chang, and Wu 2020](#)). Likewise, [Lan, Chang, and Wu \(2020\)](#) also found that the El Niño Southern Oscillation (ENSO) had an impact on catch rates near the Arabian Sea.

2.2 Stock structure

Fisheries information indicates that adult yellowfin are distributed continuously throughout the entire tropical Indian Ocean, but some more detailed analysis of fisheries data suggests that the stock structure may be more complex. The tag recoveries may indicate that the western and eastern regions of the IO support relatively discrete sub-populations of yellowfin tuna. Studies of stock structure using DNA techniques have indicated that there may be genetically discrete sub-populations of yellowfin tuna in the northwestern IO ([Dammannagoda, Hurwood, and Mather 2008](#)) and within Indian waters ([Kunal et al. 2013](#)). A recent study of stock structure using the gene sequencing technology along with a basin-scale sampling design indicated genetic differentiation between north and south of the equator within the IO, and possibly additional genetic structure within the locations north of the equator ([Grewe et al. 2020](#)). Parasite composition and abundance suggest limited movement of yellowfin between Indonesian archipelago (eastern IO) and the Maldives (central IO) ([Moore et al. 2019](#)). Isotope studies have also suggested relatively limited movement, with resident behaviour at the temporal scale of their muscle turnover

(~ 3 months) (Ménard et al. 2007). Otolith chemistry analyses concluded that fisheries operating in the western IO are mainly composed of fish with western origin, which suggest limited movement from east to west (Artetxe-Arrate et al. in review). These studies generally support the potential presence of population units of yellowfin tuna within the IO, despite that considerable uncertainty remains on sub-regional population structure in this region. This assessment assumes that the IO yellowfin tuna stock consists of several interconnected regional populations (Figure 1) that have the same biological characteristics; however, we acknowledge that more studies are needed to reveal the structure of this species.

2.3 Fisheries

Yellowfin tuna are harvested with a diverse variety of gear types, from small-scale artisanal fisheries (in the Arabian Sea, Mozambique Channel and waters around Indonesia, Sri Lanka, the Maldives, and Lakshadweep Islands) to large gillnetters (from Oman, Iran and Pakistan operating mostly but not exclusively in the Arabian Sea) and distant-water longliners and purse seiners that operate widely in equatorial and tropical waters (Figure 5). Purse seiners and gillnetters catch a wide size range of yellowfin tuna, whereas the longline fishery takes mostly adult fish (Figure 3).

Prior to 1980, annual catches of yellowfin tuna remained below about 80,000 mt and were dominated by longline catches (Figure 2). Annual catches increased markedly during the 1980s and early 1990s, mainly due to the development of the purse-seine fishery as well as an expansion of the other established fisheries (fresh-tuna longline, gillnet, baitboat, handline and, to a lesser extent, troll). A peak in catches was recorded in 1993, with catches over 400,000 mt, the increase in catch almost fully attributable to longline fleets, particularly longliners flagged in Taiwan, which reported exceptional catches of yellowfin tuna in the Arabian Sea. The Taiwanese longline fishery in the IO has been equipped with super-cold storage. Since around 1986, the fleet has fished more frequently with deep sets.

Catches declined in 1994, to about 350,000 mt, remaining at that level for the next decade then increasing sharply to reach a peak of about 520,000 mt in 2004-2005 driven by a large increase in catch by all fisheries, especially the purse-seine (free school) fishery. Total annual catches declined sharply from 2004 to 2007 and remained at about 300,000 mt during 2007–2011. In 2012, total catches increased to about 400,000 mt and were maintained at about that level through 2013 to 2015. Total catches increased to an average of 430,000 mt between 2016 and 2019, and a maximum of close to 450,000 mt in 2019 (Figure 2), despite IOTC Resolution 17/01 which requested major fleets to substantially reduce their yellowfin catches below the 2014 or 2015 catch level. Furthermore, catch levels of about 440,000 mt reported for 2018 might be under-estimated (to some extent) because of changes in data processing methodology by European Union-Spain for its purse seine fleet for that year (IOTC 2021).

In recent years (2015–2023), purse seine has been the dominant fishing method harvesting 36% of the total IO yellowfin tuna catch (by weight), with the gillnet and handline fisheries, comprising 20% and 18% of the catch, respectively. There was a substantial increase in the catch by handline in 2020 (Figure 2). A smaller component of the catch was taken by industrial longline (5%), and the regionally important baitboat (4%) and troll (4%) fisheries. The recent increase in the total catch has been mostly attributable to an increase in catch from the gillnet and handline fisheries.

The purse-seine catch is generally distributed equally between free-school and associated (log and FAD sets) schools, although the large catches in 2003–2005 were dominated by fishing on

free-schools. Conversely, during 2015–2023 the purse-seine catch was dominated (70%) by the associated fishery.

Historically, most of the yellowfin catch has been taken from the western equatorial region of the IO (44%; region 1b, Figure 5) and, to a lesser extent, the Arabian Sea (26%), the eastern equatorial region (24%, region 4) and the Mozambique Channel (5%; region 2). The purse-seine and baitboat fisheries operate almost exclusively within the western equatorial region, while catches from the Arabian Sea are principally by handline, gillnet, and longline (see Figure 5). Catches from the eastern equatorial region (region 4) were dominated by longline and gillnet (around Sri Lanka and Indonesia). The southern IO (region 3) accounts for a small proportion of the total yellowfin catch (1%) taken exclusively by longline.

In recent years (2008–2012), due to the threat of piracy, the bulk of the industrial purse seine and longline fleets moved out of the western waters of Region 1b to avoid the coastal and off-shore waters off Somalia, Kenya and Tanzania. The threat of piracy particularly affected the freezer longline fleet and levels of effort and catch decreased markedly from 2007. The total catch by freezing longliners declined to about 2,000 mt in 2010, a 10-fold decrease in catch from the years before the onset of piracy. Purse seine catches also dropped in 2007–2009 and then started to recover. Piracy off the Somali coast was almost eliminated by 2013 but longline catches have not recovered.

The sizes caught in the IO range from 30 cm to 180 cm fork length (Figure 3). Intermediate age yellowfin are seldom taken in the industrial fisheries, but are abundant in some artisanal fisheries, mainly in the Arabian Sea. Newly recruited fish are primarily caught by the purse seine fishery on floating objects and the pole-and-line fishery in the Maldives. Males are predominant in the catches of larger fish at sizes larger than 150 cm (this is also the case in other oceans). Medium sized yellowfin concentrate for feeding in the Arabian Sea.

3 Model structure

3.1 Spatial stratification

The geographic area considered in the assessment is the IO, defined by the coordinates 40°S-25°N and 20°E-150°E. Earlier yellowfin stock assessments have adopted a five-area spatial structure ([Langley, Herrera, and Million 2012](#)), but several issues were identified for that structure. Since 2015, a four-area spatial structure is used for this stock (Figure 1). The Arabian Sea (area 1a) and western equatorial region (area 1b) make up the model area 1 but kept the fishery information separated (i.e., areas-as-fleet approach) to account for differences in selectivity between these sub-areas ([Punt 2019](#)). The spatial structure retains two regions that encompass the main year-round fisheries in the tropical area (areas 1 and 4) and two austral, subtropical regions where the longline fisheries occur more seasonally (areas 2 and 3).

The current spatial structure separates the purse-seine fishery in the northern Mozambique Channel (10-15°S) from the equatorial region, as the fishery in the northern Mozambique Channel exhibits strong seasonal variation in effort and operates differently from the equatorial region ([Langley 2015](#)). There is also a separation of the purse-seine fishery between the western and eastern tropical region with the current boundary between region 1b and region 4. In addition to the four-area configuration, we also evaluated two more spatial structures: one-area and two-area configurations (see more details in Appendix XX). The 2021 assessment also evaluated a

modified version of the four-area structure (Fu et al. 2021), but there were some constraints to evaluate that configuration in the current assessment (see Section XX).

3.2 Temporal stratification

The time period covered by the assessment is 1950–2023, which represents the period for which catch data are available from the commercial fishing fleets. Langley (2015) suggested that the assessment results were not sensitive to the early catches from the model (pre-1972) and commencing the model in 1950 or 1972 (assuming unexploited equilibrium conditions) yielded very similar results.

The time step in the assessment model was quarter (i.e., three months duration, four quarters per year), representing a total of 296 model time steps. The definition of these time steps enabled recruitment to be estimated for each quarter to approximate the continuous recruitment of yellowfin in the equatorial regions. In addition, the quarterly model time step precluded the estimation of seasonal model parameters, particularly the movement parameters. Fu et al. (2018) explored an alternative annual/seasonal model structure which explicitly estimated seasonal movement dynamics. However, the alternative temporal structure did not yield substantially different results.

4 Model inputs

Catch (1950-2023) and size (1952-2023) information was provided by the IOTC Secretariat in a comma-separated values (CSV) format. These datasets and the metadata can be found online at the IOTC website: <https://iotc.org/documents/WPTT/26AS/Data/01>. Four indices of abundance were also available: joint longline, purse seine fishing on free schools, purse seine fishing on associated schools, and the associative behavior-based abundance index. These indices can also be found online at: <https://iotc.org/documents/standardised-cpue-yft-and-bet>. Release and recovery data (2005-2014) from two tagging programs were also available, as well as age data from the GERUNDIO project (2013-2021).

4.1 Definition of fisheries

The assessment adopted the equivalent fisheries definitions used in the previous stock assessments (Table 2). First, nine *fishery groups* were defined based on fleet, gear, purse seine set type, and type of vessel in the case of longline fleet (Table 2), which represent relatively homogeneous fishing units, with similar selectivity and catchability characteristics that do not vary greatly over time. Then, *fishery groups* were divided into regions (Table 3), producing a total of twenty-one *fisheries* in the assessment model. A brief description of each *fishery group* is provided below.

The longline fishery was partitioned into two main components:

- *Freezing longline fisheries (LL)*, or all those using drifting longlines for which one or more of the following three conditions apply: (i) the vessel hull is made up of steel; (ii) vessel length overall of 30 m or greater; (iii) the majority of the catches of target species are preserved frozen or deep-frozen. A composite longline fishery was defined in each model

area aggregating the longline catch from all freezing longline fleets (principally Japan and Taiwan).

- *Fresh-tuna longline fisheries (LF)*, or all those using drifting longlines and made of vessels (i) having fibreglass, fibre reinforced plastic, or wooden hull; (ii) having length overall less than 30 m; (iii) preserving the catches of target species fresh or in refrigerated seawater. A composite longline fishery was defined aggregating the longline catch from all fresh-tuna longline fleets (principally Indonesia and Taiwan) in region 4, which is where the majority of the fresh-tuna longliners have traditionally operated.
- The purse-seine catch and effort data were apportioned into two separate method fisheries: catches from sets on associated schools of tuna (log and drifting FAD sets; *LS*) and from sets on unassociated schools (free schools; *FS*).
- A single baitboat fishery (*BB*) was defined within region 1b (essentially the Maldives fishery).
- Gillnet fisheries (*GI*) were defined in the Arabian Sea (region 1a), including catches by Iran, Pakistan, and Oman, and in region 4 (Sri Lanka and Indonesia).
- Three troll fisheries (*TR*) were defined, representing separate fisheries in regions 1b (Maldives), 2 (Comoros and Madagascar) and 4 (Sri Lanka and Indonesia).
- A handline fishery (*HD*) was defined within region 1a, principally representing catches by the Yemeni fleet.
- A miscellaneous “Other” fishery (*OT*) was defined comprising catches from artisanal fisheries other than those specified above (e.g. trawlers, small purse seines or seine nets, sport fishing, and a range of small gears).

4.2 Catch

The catch datasets was composed of information about time (year and month), fleet, gear type, type of association of the fish school, grid code at a $5^\circ \times 5^\circ$ resolution, and catch in weight (metric tons) and numbers. The grid code contained information on the grid resolution, quadrant, and longitude and latitude of the corner of the grid.

4.2.1 Processing

We followed the next steps to produce the catch input for SS3:

- Month information was used to assign quarters (i.e., four quarters from January to December).
- Fishery group was assigned based on the fleet, gear type, and the type of association of the fish school.
- Catch was summed by year, quarter, fleet, gear type, fishery group, and grid code.
- The grid code was used to calculate the longitude and latitude of the center of the grid (called *centroid* hereafter).
- The centroid was used to assign regions used in the assessment model (Figure 9).
- Fishery was assigned based on fishery group and region.

4.2.2 Reassignment

To simplify the fishery structure in the stock assessment model, we reassigned catches in areas with low fishing activity to main areas as done in the 2021 assessment.

- *LF* fisheries: catch in regions 1 to 3, representing only ~ 3% of the total catches over the time series, were assigned to region 4.
- *FS* and *LS* fisheries: purse seine catches in region 1a and 3 were reassigned to region 1b and 4, respectively.
- *BB* fisheries: a small proportion of the total baitboat catch and effort occurs on the periphery of region 1b, within regions 1a and 4. Therefore, we assigned all *BB* catches to region 1b.
- *GI* fisheries: a very small proportion of the total gillnet catch and effort occurs in regions 1b and 2, which was reassigned to area 1a. Likewise, catch in region 3 was reassigned to region 4.
- *TR* fisheries: moderate troll catches are taken in regions 1a and 3, which were reassigned to regions 1b and 4, respectively.
- *HD* fisheries: moderate handline catches are taken in regions 1b, 2 and 4, which were reassigned to region 1a.
- *OT* fisheries: catch from region 1b and 2 was reassigned to region 1a, while catch from region 3 was reassigned to region 4.

4.2.3 Aggregation

After catch reassignment, catch data (in metric tons) was summed by year, quarter, and fishery, and then organized in a SS3 format assuming an error of 0.01. Overall, the time series of catches were quite similar to the catch series included in the 2021 assessment (Figure 11). The largest differences were observed for *OT* and *TR* in region 4, especially during the last two decades. Also, current catch estimates for *LL* in region 3 are slightly larger than the previous assessment during 2008-2020. The changes are mostly attributed to revisions of catch estimation by the IOTC Secretariat.

4.3 Size data

The size data was composed of information about time (year and month), fleet, gear type, type of association of the fish school, grid code, number of fish sampled per fork length bin (cm), and the score of reporting quality (RQ) (Herrera 2010; IOTC 2024). The length bin width was 2 cm and the length bins spanned from 10 to 340 cm. The data were collected from a variety of sampling programmes, which can be summarized as follows:

- *FS* and *LS* fisheries: Length-frequency samples from purse seiners have been collected from a variety of port sampling programmes since the mid-1980s. The samples are comprised of very large numbers of individual fish measurements. The length frequency samples are available by set type with sets catches from associated sets typically composed of smaller fish than free school catches (Figure 3). The size composition of the catch from the free-school fishery is bimodal, being comprised of the smaller size range of yellowfin and a broad

mode of larger fish (Figure 3). The bimodal distribution is likely to have reflected different types of schools in the catch composition (e.g., free schools of mostly large adult yellowfin, or mixed species schools consisting of smaller yellowfin, M. Chassot, pers. comm.). Hence, the relative composition of large ($>80\text{cm}$) vs. small ($<80\text{cm}$) yellowfin in the purse seine free schools fluctuates considerably over time. Between 2010 and 2020, there was both a dip in the average size of large fish caught in the FAD fishery, and a temporary increase in the average sizes of large fish caught in the free school fishery (Figure 4). There is also considerable catch of smaller fish taken during free school fishing operation in the Mozambique Channel area in region 2 (Chassot 2014). The free-school fishery in region 4 appears to catch larger fish (Figure 6).

- *LL* fishery: Length and weight data have been collected from sampling at ports and aboard Japanese commercial, research vessels, and observer programmes. Weight frequency data collected from the fleet have been converted to length frequency data via a processed weight-whole weight conversion factor and a weight-length key. Length frequency data from the Taiwanese longline fleet from 1980–2003 were included in the 2018 assessment, although data from the more recent years were excluded due to concerns regarding their reliability (Greehan and Hoyle 2013). Length data are also available from other fleets (e.g., Seychelles, Korean, China, etc.) in more recent years. Analyses of size data show that the average lengths of yellowfin caught by the longline fleet are generally larger in the southern regions, particularly in the southwest (Hoyle 2021). There is considerable temporal variation in the length of fish caught (Figure 4), but some of this variation is inconsistent between datasets, such as temporal patterns of variation in the 1970s that differ between length and weight data from the Japanese fleet. For all longline fisheries there was a marked decline in the size of fish caught by Japan during the 1950s and 1960s, while the size of fish caught stabilised during the 1970s and 1980s (Figure 4).
- *LF* fishery: Length and weight data were collected in port, during unloading of catches, for several landing locations and time periods, especially on fresh-tuna longline vessels flagged in Indonesia and Taiwan/China (IOTC-OFCF sampling).
- *GI* fishery: Samples come from Iran, Pakistan, Sri Lanka, and Oman in the Arabian Sea from 1987, and from Indonesia and Sri Lanka in other tropical areas from 1975.
- *BB* fishery: Size data come principally from the Maldivian fleet from 1983 with a large proportion of juveniles. Also, samples from Indonesia and Sri Lanka are also available for some years but with a low sample size.
- *TR* fishery: Samples come mainly from Comoros in the western IO from 2015, although some small samples are also available from EU (France, Mayotte) and Maldives. In the eastern IO, size data come from Indonesia (1985-1990 and after 2019) and Sri Lanka (1994-2018) fleets.
- *HD* fishery: Samples come exclusively from the Arabian Sea region and from a high diversity of fleets, although the Maldivian fleet has been the most consistent over the years (1985-2023). Limited sampling was conducted over the last decade.
- *OT* fishery: Samples are available from 1983 in the eastern IO and from 1997 in the western IO. The main fleets are the Indonesian, Sri Lankan, Maldivian, and Mozambique. Limited samples are available during the last few years.

The IOTC Secretariat provided two size datasets with two distinct grid resolutions:

- *Original data:* the size dataset had six main types of grid dimensions (see Table 1 and Figure 8), although $\sim 97\%$ of observations were category 5 or 6. A seventh category was also present that covered the Seychelles National Jurisdiction Area, but those observations were removed from the size database. This type of size dataset was used in the 2021 assessment.
- *cwp55 data:* the size dataset was provided at a $5^\circ \times 5^\circ$ grid resolution (Figure 9).

4.3.1 Processing

First, we removed gear types with unclear classification (*HOOK*, *HATR*, *PSOB*, and *PS* with unclassified school type *UNCL*). Then, we reduced the number of length bins in the data by summing the number of sampled fish equal or larger than 198 cm and assign it to the 198 cm length bin. Likewise, we summed the number of sampled fish ≤ 10 cm and assigned to the 10 cm length bin. We followed these steps to produce the SS3 size inputs:

- Month information was used to assign quarters, having four quarters from January to December.
- Fishery group was assigned based on the fleet, gear type, and type of association of the fish school.
- The number of sampled fish per length bin was summed and the RQ was averaged by year, quarter, fleet, gear type, fishery group, and grid code.
- The grid code was used to calculate the grid centroid.
- The centroid was used to assign regions. Note that this region assignment varied depending on the type of dataset (see Figure 8 and Figure 9).
- Fishery was assigned based on fishery group and region.
- We converted the length bin width from 2 to 4 cm. To do so, we summed the number of sampled fish from pairs of length bins (e.g., 10 and 12 cm were summed and assigned to 10 cm, 14 and 16 cm were summed and assigned to 14 cm, and so on). After this conversion, we had a total of 48 length bins.

Only for the size dataset with regular grids, we then assigned the catch (in numbers) that corresponded to every observation in the size data (i.e., year, quarter, grid, fleet, and gear type). We found a perfect match for $\sim 74\%$ of cases but there were some size observations without catch. In order to fill in these catch gaps, we followed an imputation procedure with five levels:

- *Level 1:* Fill in catch gaps with the average catch per grid for a given year, quarter, fleet, gear type, and fishery group.
- *Level 2:* Fill in catch gaps with the average catch per grid for a given year, fleet, gear type, and fishery group.
- *Level 3:* Fill in catch gaps with the average catch per grid for a given year, gear type, and fishery group.
- *Level 4:* Fill in catch gaps with the average catch per grid for a given gear type and fishery group.
- *Level 5:* Fill in catch gaps with the average catch per grid for a given fishery group.

Figure 10 shows the percentage of size observations that needed each level of imputation.

4.3.2 Reassignment

We conducted the size data reassignment as done for the catch data in Section 4.2.2.

4.3.3 Filtering

In order to remove inconsistent patterns in the length frequency data, we carried out these filters:

- The first filter was to remove observations with less than 100 fish sampled and not considered best quality based on the RQ score.
- *LL* fishery: a review of the longline size data shows that the sampling behaviour of Taiwanese and Seychelles fleets (mostly reflagged Taiwanese vessels) have changed over time, with patterns in the logbook length data inconsistent with other fleet ([Hoyle 2021](#)), and as such the WPTT23 (Data Preparatory) recommended omitting all Taiwanese and Seychelles logbook length data from the 2021 assessment ([IOTC 2021](#)). Following this advice, we removed length frequency data from the Taiwanese and Seychelles longline logbooks from the final length frequency data sets.
- *LL* fishery: longline length frequency data during 1970-1995 and 2010-2020 in region 1a was removed.
- *LL* fishery: attempts to fit the size data of this fishery in past assessments suggested that the large decline in mean size observed before 1960 is inconsistent with the yellowfin population dynamics. Hoyle ([2021](#)) suggest that selectivity may have changed during this early period and recommend avoiding fitting to these data with the same selectivity. Therefore, we omitted longline size data before 1960 for regions 1b, 2, 3, and 4.
- *LL* fishery: longline length frequency data in 2001-2005, 2015, and 2019 in region 4 was removed.
- *LF* fishery: we removed size data before 2005.
- *GI* fishery: we removed size data from the Sri Lankan fleet in 2021.
- *HD* fishery: we removed size data from the Maldivian fleet in 2003.
- *OT* fishery: we removed size data sampled during 2021-2022 in region 1a, and data sampled in 2016 in region 4.
- *TR* fishery: we removed size data in region 1b and 2. Also, we removed size data in region 4 from 2016 to 2019.

The filters applied to the *LL*, *LF*, and *TR* fisheries were also applied in the 2021 assessment. Other filters were exclusive to the current assessment.

4.3.4 Aggregation

After filtering and in order to aggregate the size data by year, quarter, and fishery for SS3, we followed two approaches: simple and catch-raised aggregation.

- *Simple aggregation*: this type of aggregation was performed only for the *original* size dataset. We summed the number of sampled fish per length bin and averaged the RQ values by year, quarter, and fishery. RQ values was treated as input sample size in SS3, but we also tested using an input sample size of 5 for every observations as done in the 2021 assessment. This aggregation approach was used in the 2021 assessment and assumes

that the collection of samples was broadly representative of the operation of the fishery in each quarter.

- *Catch-raised aggregation*: this type of aggregation was performed only for the *cwp55* size dataset. We performed a catch-weighted sum of the number of sampled fish by length bin and a catch-weighted average of the RQ values. RQ values was treated as input sample size in SS3.

A graphical representation of the availability of length samples is provided in Figure 7. The longline fishery provided size data from the 1960s but with relatively low quality during the first decades. Most of the size information started to appear in the 1980s, being the purse seine fisheries the most consistent and with the highest quality (Figure 7). The RQ scores did not differ between aggregation methods.

The differences in size compositions between the simple and catch-raised aggregation methods were minimal for most fisheries (Figure 3). The largest differences were found for the *OT* fisheries in region 1a and 4, and for the handline fishery in region 1a. We observed an increase in the mean length for the *LL* fisheries in all regions (Figure 4). In the case of the free school purse seine fishery, we also noted an increase in mean length over time in region 1b. Conversely, we noted a decrease in the mean length for the log school purse seine fishery over the years in regions 1b, 2, and 4, especially from 1980 to \sim 2005. For the handline fishery, we noted an increase in mean length from the 1990 until \sim 2010, and then a decrease until recent years. These patterns were quite similar between the two aggregation methods (Figure 4).

The size compositions used in the current assessment were comparable with the 2021 assessment (both using the simple aggregation approach), although small differences can be observed for the *OT* and *TR* fisheries in region 4 (Figure 15). Regarding mean length, most fisheries other than longline had similar tendencies over time when comparing 2021 and current values. For longline *LL* fisheries, size compositions in 2021 had larger mean lengths before 1990 compared with the current size compositions (Figure 16).

4.4 Indices

4.4.1 Longline CPUE

Standardised CPUE indices (1975-2023) were derived using a hurdle generalized linear model (GLM) from longline catch and effort information provided by Japan, Korea, and Taiwan (Matsumoto et al. 2024). The data used for the standardization included operation date, fishing location, vessel ID, fishing effort (number of hooks per set), and catch in numbers. Cluster analyses of species composition data for each fleet and model area were used to separate datasets into fisheries understood to target different species. Selected clusters were then combined and standardized using GLMs. The log-transformed yellowfin catch per number of hooks set was the dependent variable of the positive model component, while the probability of catch rate being zero was the dependent variable in the binomial model component. In addition to the year and quarter variables, GLMs included covariates for 5° square location, cluster, and vessels ID. The data used in the GLM was subsampled from the operational data ($\$10 - 30^\circ \times \5° grid). Moreover, data from regions 1a and 1b were combined as a single region. The CPUE indices was produced at an annual and quarterly time step and for each model area.

The CPUE index produced for region 1 was assigned to area 1b in the stock assessment model. For the regional longline fisheries, a common catchability coefficient (and selectivity) was estimated in the assessment model, thereby, linking the respective CPUE indices among regions. This significantly increases the power of the model to estimate the relative (and absolute) level of biomass among regions. However, as CPUE indices are essentially density estimates it is necessary to scale the CPUE indices to account for the relative abundance of the stock among regions. For example, a relatively small region with a very high average catch rate may have a lower level of total biomass than a large region with a moderate level of CPUE.

We determined regional scaling factors that incorporated both the size of the region and the relative catch rate to estimate the relative level of exploitable longline biomass among regions. This approach was also used in the 2021 assessment, and is similar to that used in the Western and Central Pacific Fisheries Commission (WCPFC) regionally disaggregated tuna assessments. Hoyle (2018) proposed a set of regional weighing factors for IO yellowfin based on aggregated longline catch effort data. The authors recommended the estimates by method ‘8’ for the period 1979–1994 (referred to as ‘7994m8’, see Table 2 of Hoyle (2018)) to be included in the current assessment. The relative scaling factors calculated for regions 1–4 are 1.674, 0.623, 0.455 and 1, respectively.

For each of the principal longline fisheries, the standardized CPUE index was normalized to the mean of the period for which the region scaling factors were derived (i.e., the GLM index from 1979–1994). The normalized GLM index was then scaled by the respective regional scaling factor to account for the regional differences in the relative level of exploitable longline biomass among regions (Figure 12).

A number of important trends are evident in the CPUE indices:

- The western tropical (region 1b) CPUE increased during the late 1970s and early 1980s, then suddenly declined from 1987 to 1990. After 1990, the CPUE in this region remained roughly stable until the late 2000s, that coincided with a number of piracy incidents in the western Indian Ocean (2008–2011). After that time, it remained close to the lowest level observed in that region but showed very large seasonal and annual variations. From 2020, we noticed a remarkable increase in CPUE values compared to the 2013–2019 period.
- The eastern tropical region 4 followed a similar pattern until 1990 but then declined steadily, and by 2016 was also close to the lowest level in the time series. The recent decline in CPUE in this region is consistent with a decline in the proportion of yellowfin in the combined tuna catch from the Japanese longline fleet in the eastern IO. It is unclear whether the change in species proportion is related to a decline in the abundance of yellowfin in the region (relative to the other species) or a regional change in the targeting of the fishing fleet. However, there is an indication that there has been a differential shift towards deeper longline gear (greater HBF) in the eastern IO since 2000 and this may indicate a shift in targeting toward bigeye tuna in this region (Hoyle pers. comm. additional JP LL analyses). Such factors may not be adequately accounted for in the standardisation of the yellowfin CPUE data.
- The CPUE indices in western temperate region 2 followed a similar pattern to the western tropical indices, with a decline until the late-1970s followed by an increase until the late 1980s, and subsequently a slow decline with significant variability. From 1990, the CPUE in this region has remained roughly stable, with no remarkable increase during the last years.

- The CPUE indices from region 3 are low compared to the other three regions reflecting the low regional scaling factor. However, the overall trend in the CPUE indices is broadly comparable to the other regions. The eastern temperate region 3 the pattern was similar to the western temperate area before 1979. After 1979 catch rates increased until the mid-2000's, but then declined rapidly and reached their lowest observed levels by 2016. In this region, we also noticed an increase in CPUE for the last years.
- There is an exceptionally high peak in CPUE indices 1976–78, which is also associated with a high uncertainty. Hoyle, Satoh, and Matsumoto (2017) showed that this discontinuity exists in Japanese, Taiwanese and Korean data, and in multiple regions in multiple oceans, and for both bigeye and yellowfin tuna. Hoyle, Satoh, and Matsumoto (2017) suggested this is unlikely to be explained by changes to the population or catchability but may be associated with catch reporting and data management.
- The spike in the CPUE indices around 2012 in the west equatorial region (region 1) was evident for most fishing fleets. Several hypotheses have been proposed on what could have caused CPUE to have increased, including a return to fishing in areas that were most affected by piracy. However, further investigation is required.

The values and trend of LL CPUE used in the 2021 assessment and in this assessment were quite similar for all model areas before 1990 (Figure 14). After 1990, we noted large differences for model area 1b, where the current CPUE showed consistently larger values ($\sim 40\%$) than the 2021 CPUE, especially after 2005. For region 4, we also observed slightly larger 2024 CPUE values compared to the 2021 CPUE after 2005. However, this discrepancy was minimal for regions 2 and 3.

4.4.2 Purse seine CPUE indices

The European and associated flags purse seine fishing activities in the Indian Ocean during 1981–2020 have been monitored through the collection of logbooks and observer sampling. Standardised indices of the biomass of yellowfin caught by European purse seiners (Spain and France) from sets on free swimming schools (1991 – 2022) and sets on associated tuna schools (FOB, 2010 – 2022) were developed (Figure 13). The free school index was based on the application of a general additive mixed effect model with three components to model (Kaplan et al. 2024): i) the detection rate of free swimming schools per unit search time, ii) the probability that adult yellowfin are present in a set, and iii) the adult biomass per set given presence assuming a log-normal distribution. The FOB index was based on the application of two models: a generalized linear mixed model and a spatiotemporal model, both using a hurdle approach (Correa et al. 2024). These standardizations considered a comprehensive list of candidate covariates, including the effect of the technological improvement related to the use of echosounder buoys and environmental variables. The predicted CPUE over time was obtained using the *predict-then-aggregate* approach, which is considered best practice (Hoyle et al. 2024).

The FOB mainly informs the biomass of juvenile yellowfin, while the free-school index informs the biomass of the adult portion of the population in region 1b. The FOB index displays juvenile biomass fluctuations over the years, with larger values during 2013 and 2014, and a remarkable increase after 2020. On the other hand, the free school CPUE index showed a increase from the late 1990s until 2004, and then a dramatic decrease until 2009. From 2015, the free school index showed another dramatic decrease until 2018, and a slight recovery after that.

We evaluated the impact of incorporating the free school and FOB CPUE series as auxiliary indices (i.e., always in conjunction with the LL CPUE index) in the assessment model. In order to incorporate the free school index, we revised the region/fishery structure in which purse seine fisheries are separated into the small and large fish components (EXPAND THIS).

4.4.3 Effort creep

It is well recognised that the relationship between PS CPUE and abundance is unlikely be proportional, as the improvement of catch efficiency due to technology development is difficult to quantify, and the changes in catchability are not fully accounted for in the standardisation process. Effort creep can be defined as an unquantified increase in the average fishing power over time that disturbs the relationship of proportionality between the index and the stock trajectory (Hoyle 2024). These changes in catchability over time can affect CPUE indices, and therefore the outcomes of stock assessments. This is especially important for assessments that lack abundance indices from fishery-independent surveys, which includes the majority of the fisheries managed by tuna regional fishery management organizations (RFMOs).

For the case of longline fleets, technological advances include electronic devices to help navigate, communicate, and find target species. Synthetic materials allowed fishers to improve hooks and lines which increased probabilities of both hooking and landing. Satellite imagery improved search efficiency. Freezers increased the proportion of time spent on fishing grounds, while equipment for faster longline retrieval increased hooks set without affecting soak time (Hoyle 2024). In the 2021 yellowfin assessment in the IO, a sensitivity analysis was run during the WPTT meeting that included 1% effort creep per year for the entire period of the index, which resulted in changes in the stock depletion level. Based on the recommendations of Hoyle (2024), we evaluated two different levels of effort creep for the LL CPUE index in the current assessment: 0.5 and 1.5% per annum, which is associated with vessel turnover.

The WCPFC assessments have often estimated substantial changes in PS FAD-associated fisheries (e.g., McKechnie, Pilling, and Hampton (2017)). Using a similar approach, Kolody (2018) estimated a catchability increase of approximately 1.25% per year for the standardised purse seine effort for yellowfin from sets on associated schools. Likewise, studies on the French fleet indicate a 10% increase in catch per set associated with echosounder use, equivalent to about 1% per annum, and a 1.7 – 4.0 % increase in efficiency (stable across time) arising from fishing their own floating objects (Wain et al. 2021). Based on the recommendations of Hoyle (2024), we evaluated two different levels of effort creep for both purse seine indices in the current assessment: 1.5 and 4.35% per annum.

4.5 Age data

Age and size information was available for fish sampled between 2009 and 2022 (Figure 17) from the *GERUNDIO* project that aimed the collection and analysis of biological samples of tropical tunas, swordfish, and blue sharks to improve age, growth, and reproduction data for the IOTC. In a first step, J. Farley et al. (2021) presented this data that contained information from otoliths from 253 yellowfin tuna sampled mainly in the western IO. Then, J. H. Farley et al. (2023) presented an updated dataset with age and length information from additional 136 individuals. To calculate decimal age of sampled fish, daily and annual aging methods were used. Decimal age was calculated for each fish with an annual count based on the method developed

for yellowfin and bigeye tuna (*Thunnus obesus*) in the western Pacific Ocean (J. Farley et al. 2020). To find more details on the ageing estimation method, see J. H. Farley et al. (2023).

Age and fork length information from a total of 389 individuals was provided to be used in the current stock assessment. This source of information can be included in SS3 as conditional age-at-length data, which is important to inform growth and stock age structure (Lee et al. 2019). We followed these steps to produce the input CAAL data for SS3:

- We first removed observations with incomplete age and fleet data, retaining 375 observations.
- Month information was used to assign quarters.
- There was no discrimination between the set type for the purse seine fishery. So we assigned observations ≤ 80 cm as *LS* (log school) and > 80 cm as *FS* free school.
- Fork length was grouped into the length bins used in the assessment model.
- Fish older than 7 years was grouped into a single group (age 7).
- Region was assigned based on geographical location.
- There were several observations (~ 37%) with no geographical location, which were assigned to region 1b.

Figure 18 shows the CAAL information included in the assessment model per fishery. Adults were mostly sampled from the *LL*, *FS*, and *HD* fisheries. Conversely, juveniles were mainly present in the *LS* fishery. The lengths ranged from 18 to 182 cm and all the model ages (i.e., quarters) were sampled. The largest and oldest fish were observed in the *GI*, *LL*, and *FS* fisheries. We assumed no ageing error in the assessment model.

4.6 Tagging

Tagging data was available for inclusion in the assessment model, which consisted of yellowfin tuna tag releases and returns from the Indian Ocean Tuna Tagging Programme (IOTTP) and from the main phase of the Regional Tuna Tagging Project-Indian Ocean (RTTP-IO) conducted during 2005–2009. The IOTC has compiled all the release and recovery data from the RTTP-IO and the complementary small-scale programmes in a single database. A total of 54688 yellowfin tuna were released by the RTTP-IO program. Most of the tag releases occurred within the western equatorial region (region 1b) and a high proportion of these releases occurred in the second and third quarters of 2006 (Figure XX). Limited tagging also occurred within regions 1a and 2. The model included all tag recoveries up to the end of 2014 and there were no further recoveries since the last assessment. The spatial distributions of tag releases and recoveries are presented in Figure XX.

A total of 9916 tag recoveries (after removing tags with unknown recovery date or length) could be assigned to the fisheries included in the assessment model. Almost all of the tags released in region 1 were recovered in the home region, although some recoveries occurred in adjacent regions, particularly in region 2. A small number of tags were recovered in region 4 (from tags released in region 1b) and there were no tags recovered from region 3 (Table XX). Most of the tag recoveries occurred between mid-2006 and mid 2008 (Figure XX). The number of tag recoveries started to attenuate in 2009 although small numbers of tags were recovered up to the end of 2014.

Most of the tags were recovered by the purse seine fishery within region 1b (Figure XX). A significant proportion (35%) of the tag returns from purse seiners were not accompanied by information concerning the set type. These tag recoveries were assigned to either the free-school

or log fishery based on the expected size of fish at the time of recapture; i.e. fish larger than 80 cm at release were assumed to be recaptured by the free-school fishery; fish smaller than 80 cm at release and recaptured within 18 months at liberty were assumed to be recovered by the floating object fishery; fish smaller than 80 cm at release and recaptured after 18 months at liberty were assumed to be recovered by the free-school fishery.

For incorporation into the assessment model, tag releases were stratified by release region, time period of release (quarter) and age class. The recaptures by fishery for each release group inform the assessment model on fishing mortality and abundance and fish movement. Therefore, factors that might have affected the interpretation of tag returns need to be accounted for to minimise potential bias. Fu (2020) provides a summary of how the tag data were incorporated into the assessments of IOTC tropical tuna species, and below is a description of the procedure applied to yellowfin tuna.

4.6.1 Age assignment of tag release

The age at release was assumed based on the fish length at release and the average length-at-age from the yellowfin growth function (see Section XX). Fish aged 15 quarters and older were aggregated in a single age group. Tag releases in regions 1a and 1b were stratified in separate release groups due to the spatial separation of the individual release events. A total of 54392 releases were classified into 131 tag release groups. Most of the tag releases were in the 5–8 quarter age classes (Figure XX).

4.6.2 Initial tagging mortality

Hoyle et al. (2015) examined the effects of various covariates (e.g., individual tagger effect) on tag failures for the RTTP program and estimated a combined effect of 20% for all tropical tuna species relative to a base failure rate. No formal estimate was made for the base failure rate but a 7.5% was suggested by the WPTT in 2018 based on the assessment of the western and central Pacific tuna species. This equates to a total tag failure rate of 27.5%. For the current assessment, the number of tags in each release group was reduced by 27.5% to account for initial tag mortality.

4.6.3 Chronic tag loss

Tag recoveries were also corrected for long-term tag loss (tag shedding) based on an update of the analysis of Gaertner and Hallier (2015). Tag loss for yellowfin was estimated to be approximately 20% at 2000 days at liberty. This was accounted for through the SS3 chronological tag loss parameter (an annual rate of 0.03).

4.6.4 Reporting rate

The returns from tag release group were classified by recapture fishery and recapture time period (quarter). The results of associated tag seeding experiments conducted during 2005–2008, have revealed considerable temporal variability in tag reporting rates from the IO purse-seine fishery (Hillary et al. 2008). Reporting rates were lower in 2005 (57%) compared to 2006 and 2007 (89% and 94%). Quarter estimates were also available but was similar in magnitude (Hillary, IOTC,

and Areso 2008). This large increase over time was the result of the development of publicity campaign and tag recovery scheme raising the awareness of the stakeholders, i.e. stevedores and crew. SS3 assumes a constant fishery-specific reporting rate. To account for the temporal change in reporting rate, the number of tag returns from the purse-seine fishery in each stratum (tag group, year/quarter, and length class) were corrected using the respective estimate of the reporting rate. Following Kolody, Herrera, and Million (2011), Fu (2020), and Fu (2017), an 100% reporting rate was assumed for at-sea recoveries whereas tags recovered from Seychelles landings were corrected for reporting rates based on the quarterly estimates from Hillary, IOTC, and Areso (2008), and were also corrected for the portion of the total purse-seine catches examined for tags, based the proportions of EU purse seine catch landed in the Seychelles relative to the total EU purse seine catches (Kolody, Herrera, and Million 2011). For example, the adjusted number of observed recaptures for a LS fishery as input to the model, the reporting rate (R'_L) was calculated using the following equation:

$$R'_L = R_L^{sea} + \frac{R_L^{sez}}{P_{sez}r_{sez}}$$

where:

- R_L^{sea} is the number of observed recaptures recovered at sea for the LS fishery
- R_L^{sez} is the number of observed recaptures recovered in Seychelles for the LS fishery
- r_{sez} is the reporting rates for purse seine tags removed from the Seychelles
- P_{sez} is the scaling factor to account for the EU purse seine recaptures not landed in Seychelles

The adjusted number of recaptures for a FS fishery was calculated similarly. The SS3 reporting parameters for the purse seine fisheries were subsequently fixed at 100% in the model. Some of the other (non purse-seine) fisheries also returned a substantial number of tags. There are no direct estimates of fishery specific reporting rates for these fisheries. The reporting rates for these fisheries are estimated within the assessment model.

4.6.5 Small-scale tagging programmes

Additional tag release/recovery data are available from a number of small-scale tagging programmes. The data set included a total of 7,828 tags released during 2002-08, primarily within regions 1b (70%) and 4 (28%). A total of 366 tag recoveries were reported, predominantly from the Bait boat fishery in region 1a. There has been no comprehensive analysis of these data and there is no information available concerning the fishery specific reporting rate of these tags. The tag release/recovery data from the SS tagging programmes were not incorporated in the current range of assessment models. Earlier analysis indicated that the stock assessment results were relatively insensitive to the inclusion of these data (Langley, Herrera, and Million 2012).

Fu et al. (2018) investigated a range of alternative options for processing and incorporating the tagging data into the assessment model (see Table 5 of Fu et al. (2018)). These exploratory analyses are not repeated in the current assessment.

4.7 Environmental data

The 2018 assessment included a range of environmental data to investigate the potential for the incorporation of environmental covariates to inform the movement of fish. However, although there is evidence that there may be an association between the movement of yellowfin tuna and seasonal and temporal changes in ocean conditions in the IO, the potential relationship between environmental indices and fish movement is unclear. Langley (2016) and Fu et al. (2018) suggested that these environmental indices had no influence on the estimation of yellowfin tuna movement rates of different life stages between adjacent model regions, and seasonal variation in movement may be better accounted for by models that can explicitly incorporate seasonal effects (Fu et al. 2018). Therefore, environmental information was not included in the 2021 assessment.

A significant negative association between the Indian Ocean Dipoles (IODs) and the catch rates of yellowfin tuna was observed (Lan, Evans, and Lee 2013; Lan, Chang, and Wu 2020), as well as an impact of the El Niño Southern Oscillation (ENSO) on catch rates near the Arabian Sea (Lan, Chang, and Wu 2020). Langley, Fu, and Maunder (2023) found that periods of strong recruitment in regions 1 and 4 appear to correspond with oceanographic conditions indexed by Dipole Mode Index (DMI), but they did not find apparent correspondence between yearly trends in IO environmental conditions and the LL CPUE indices from each of the model regions. There is no strong indication that the catchability of the equatorial longline fisheries (region 1 and 4) is strongly influenced by the prevailing environmental conditions, but there is some indication that oceanographic conditions may influence short-term (1-2 yr) variation in longline catchability in the sub tropical regions (region 2 and 3) (Langley, Fu, and Maunder 2023). Variability in deviations of movement rates between region 1 and 4 is broadly consistent with the fluctuations in the DMI with higher movement estimated under positive IOD conditions.

Since no strong relationships between IO environmental conditions and the yellowfin dynamics have been identified, we did not incorporate any environmental variable in the current assessment; however, we suggest further studies on this topic.

5 Model parameters

5.1 Population dynamics

Parametrization of recruitment, growth, etc in SS.

5.2 Fishery dynamics

Selectivity principally.

5.3 Tagged fish

Tag reporting, mixing, etc.

5.4 Likelihood components

Weights, error structure, observation error, etc.

5.5 Parameter estimation and uncertainty

Hessian, max gradient, jitter analysis, likelihood profiles, etc.

5.6 Stock status

Related to MSY, depletion trends, Kobe, etc.

6 Model runs

6.1 Update from the last assessment

Describe stepwise model development, etc.

6.2 Sensitivity and structural uncertainty

Different steepness, M, etc, and spatial configurations.

7 Model results

7.1 Fits

Describe fits to data.

7.2 Parameter estimates

Describe parameter estimates.

7.3 Time series

Recruitment, SSB, etc.

7.4 Sensitivity

Describe sensitivity analyses.

8 Discussion and conclusions

Discuss. Recommendations for future assessments.

9 Acknowledgements

Write your acknowledgements.

10 Tables

Table 1: Grid size categories in the *original* size dataset.

Grid category	Resolution (latitude × longitude)
9	$30^\circ \times 30^\circ$
A	$10^\circ \times 20^\circ$
7	$10^\circ \times 10^\circ$
8	$20^\circ \times 20^\circ$
5	$1^\circ \times 1^\circ$
6	$5^\circ \times 5^\circ$
NJA_SYC	Seychelles National Jurisdiction Area

Table 2: Nine fishery groups and codes used in the current assessment.

Fishery code	Fishery group
GI	Gillnet
HD	Handline
LL	Longline
OT	Others
BB	Baitboat
FS	Purse seine, free school
LS	Purse seine, log school
TR	Troll
LF	Longline (fresh tuna)

Table 3: Fishery definition in the four-areas assessment configuration.

Fishery number	Fishery code and region
1	GI 1a
2	HD 1a
3	LL 1a
4	OT 1a
5	BB 1b
6	FS 1b
7	LL 1b
8	LS 1b
9	TR 1b
10	LL 2
11	LL 3
12	GI 4
13	LL 4
14	OT 4
15	TR 4
16	FS 2
17	LS 2
18	TR 2
19	FS 4
20	LS 4
21	LF 4

11 Figures

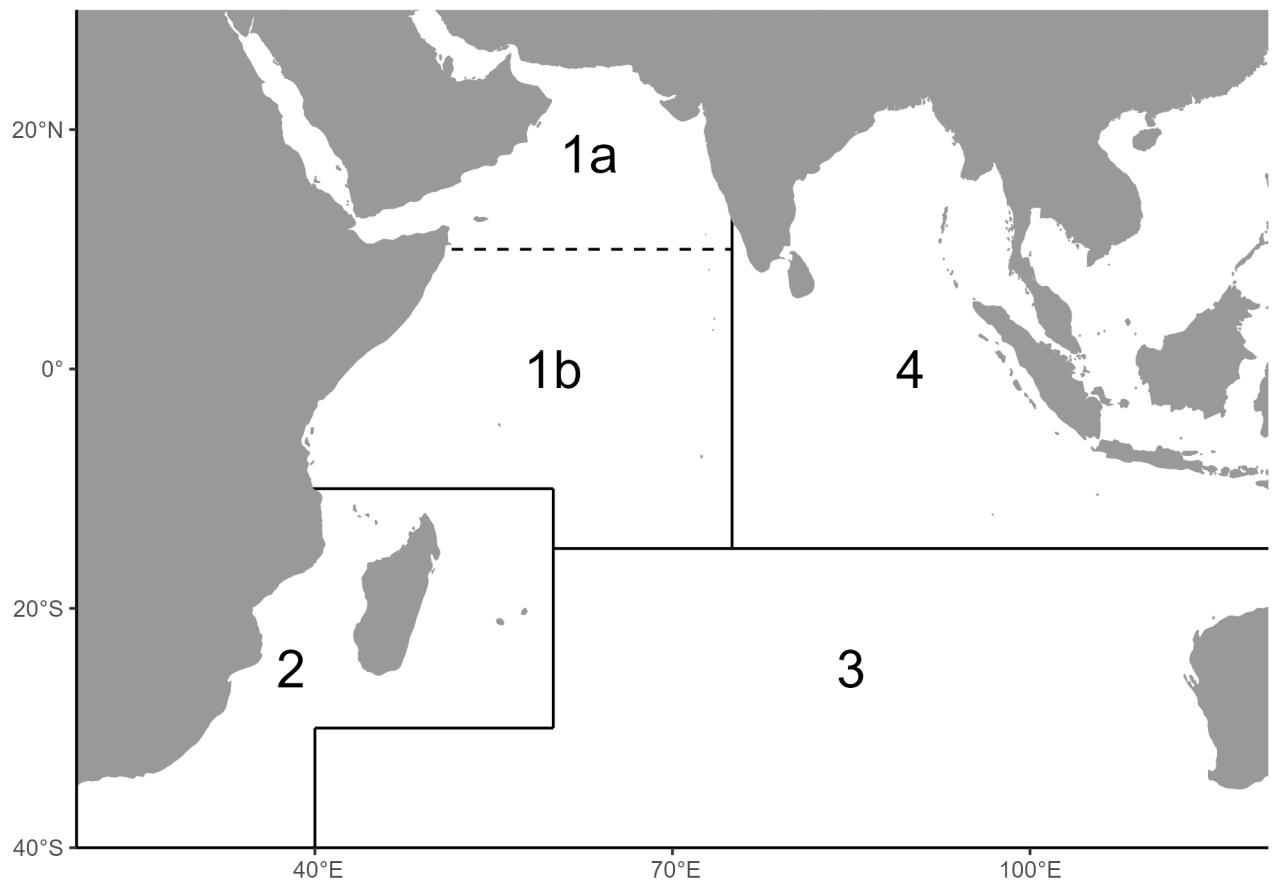


Figure 1: Model areas or regions used in the four-areas assessment configuration. The region 1 is divided into two sub-regions implicitly modelled in the assessment model using the fleets-as-areas approach.

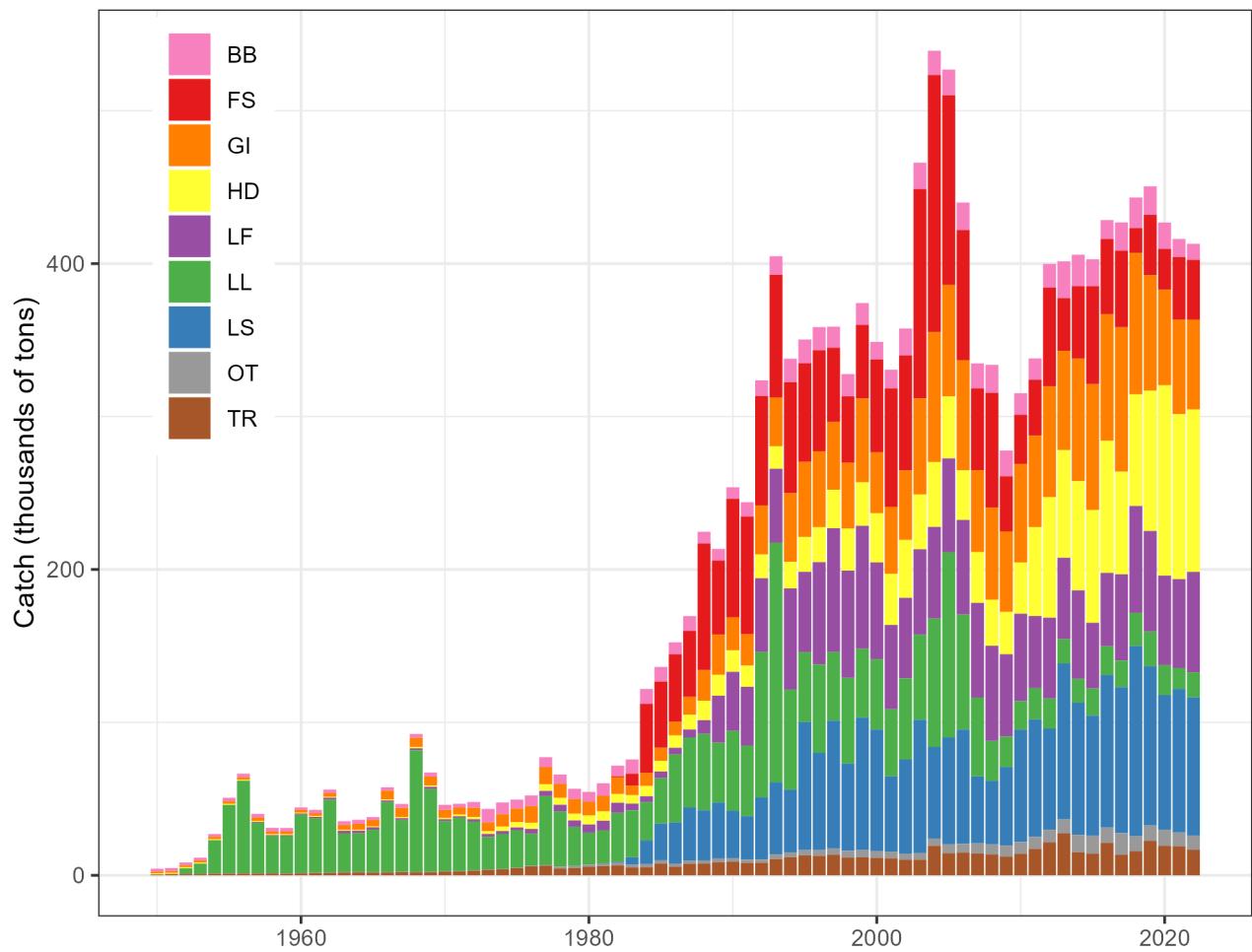


Figure 2: Total annual catch of yellowfin tuna by fishery group from 1950 to 2023. Fishery group codes are described in Table 2.

12 Appendix

12.1 Acronyms and Abbreviations

Include a table with acronyms (e.g., GLMM, SS, etc.)

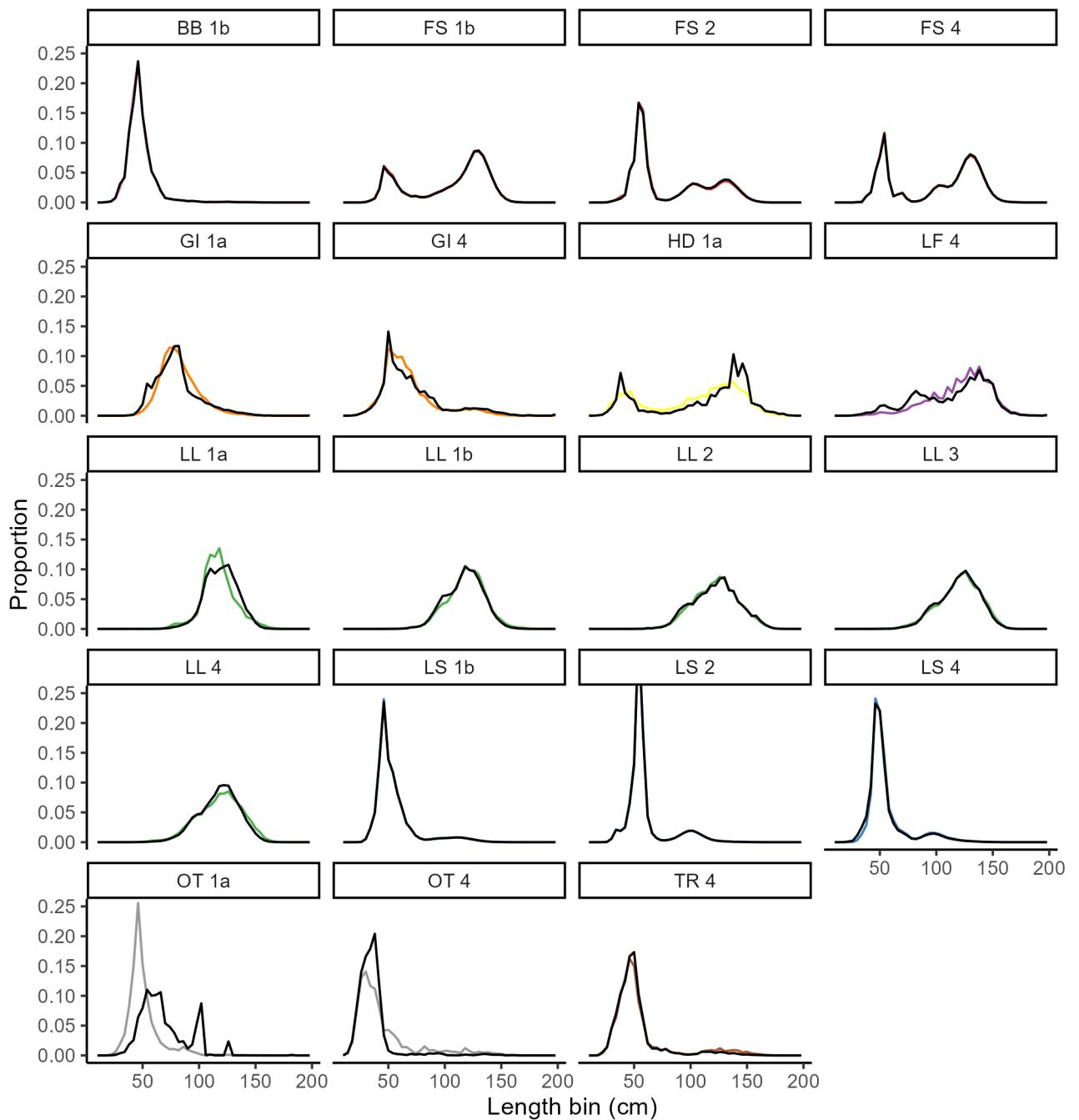


Figure 3: Size compositions per fishery included in the assessment model. Colored lines are size compositions obtained using simple aggregation while black lines used the catch-raised aggregation. Size compositions were summed over time.

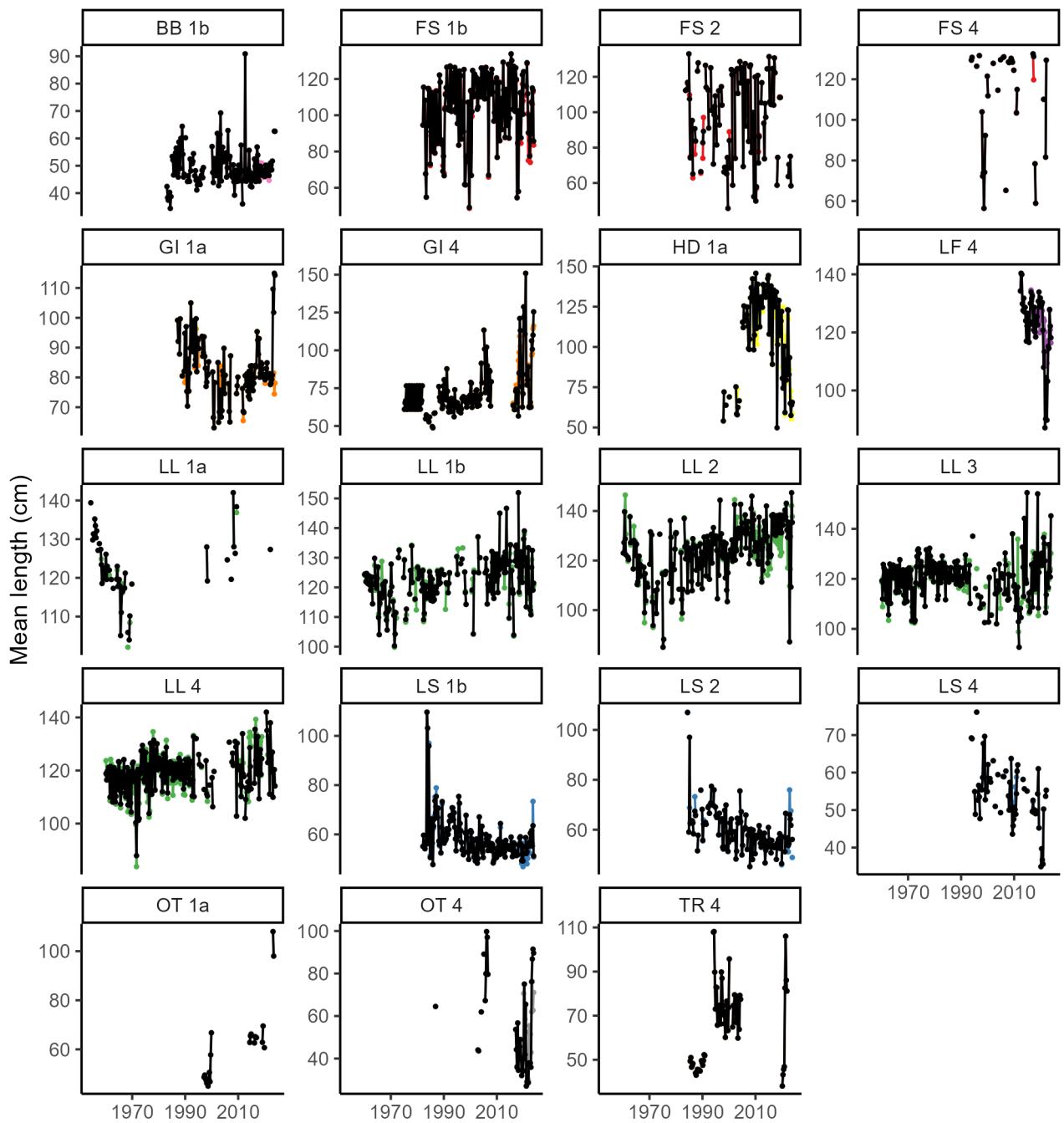


Figure 4: Mean length per quarter per fishery. Colored lines used the simple aggregation while black lines used the catch-raised aggregation.

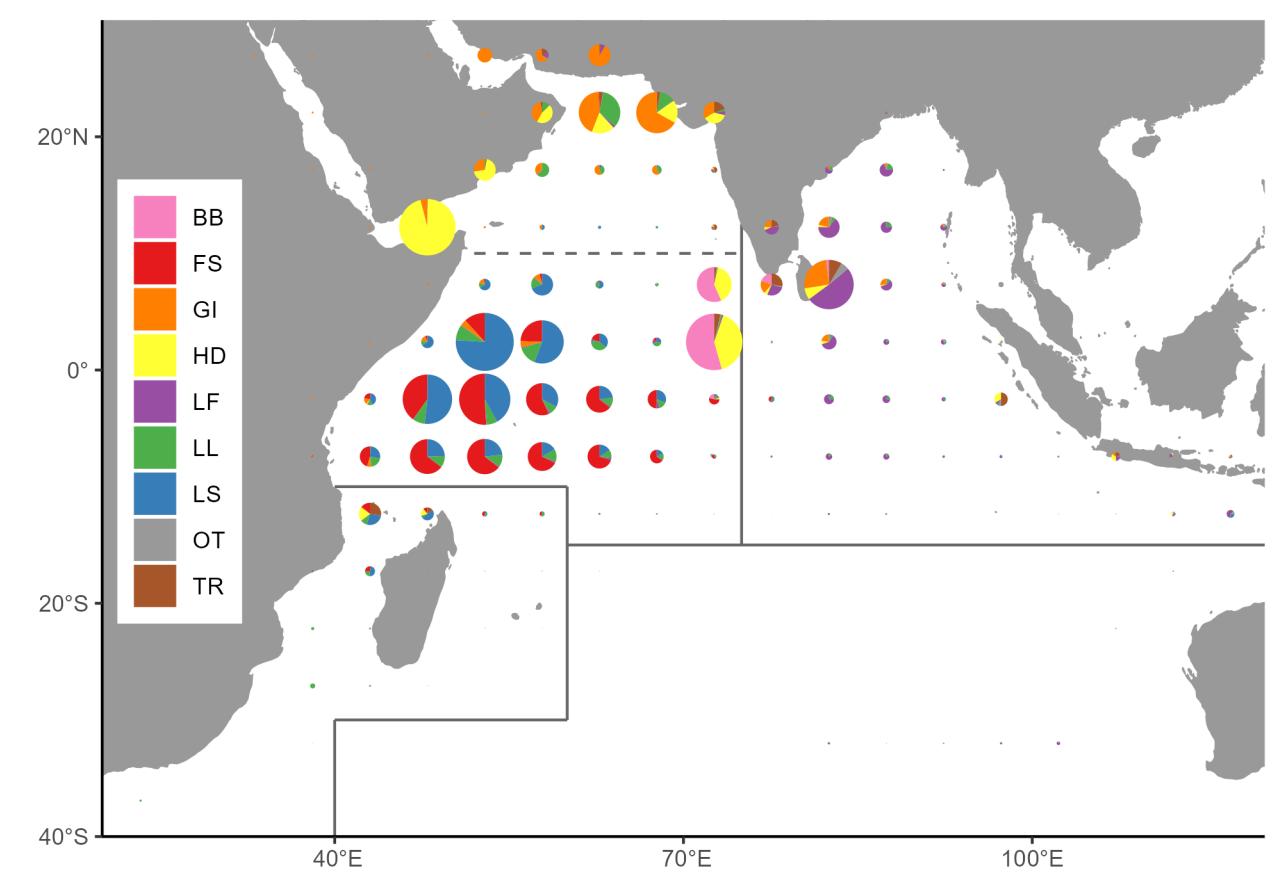


Figure 5: Spatial distribution of IO catches per fishery group. The pie radius represents the aggregated catch from 1980 to 2023. Fishery group codes are described in Table 2.

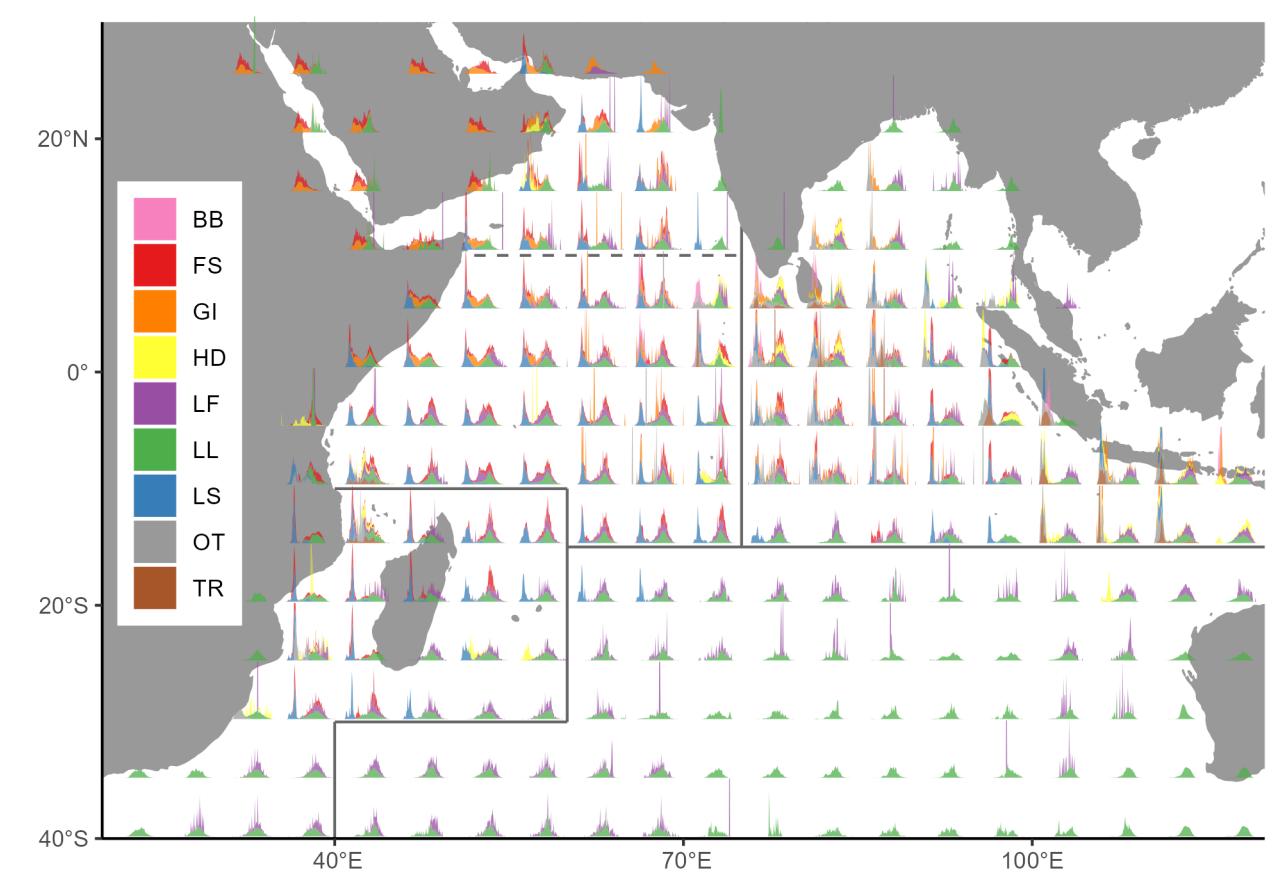


Figure 6: Spatial distribution of size compositions per fishery group. The size compositions were aggregated over time. Fishery group codes are described in Table 2.

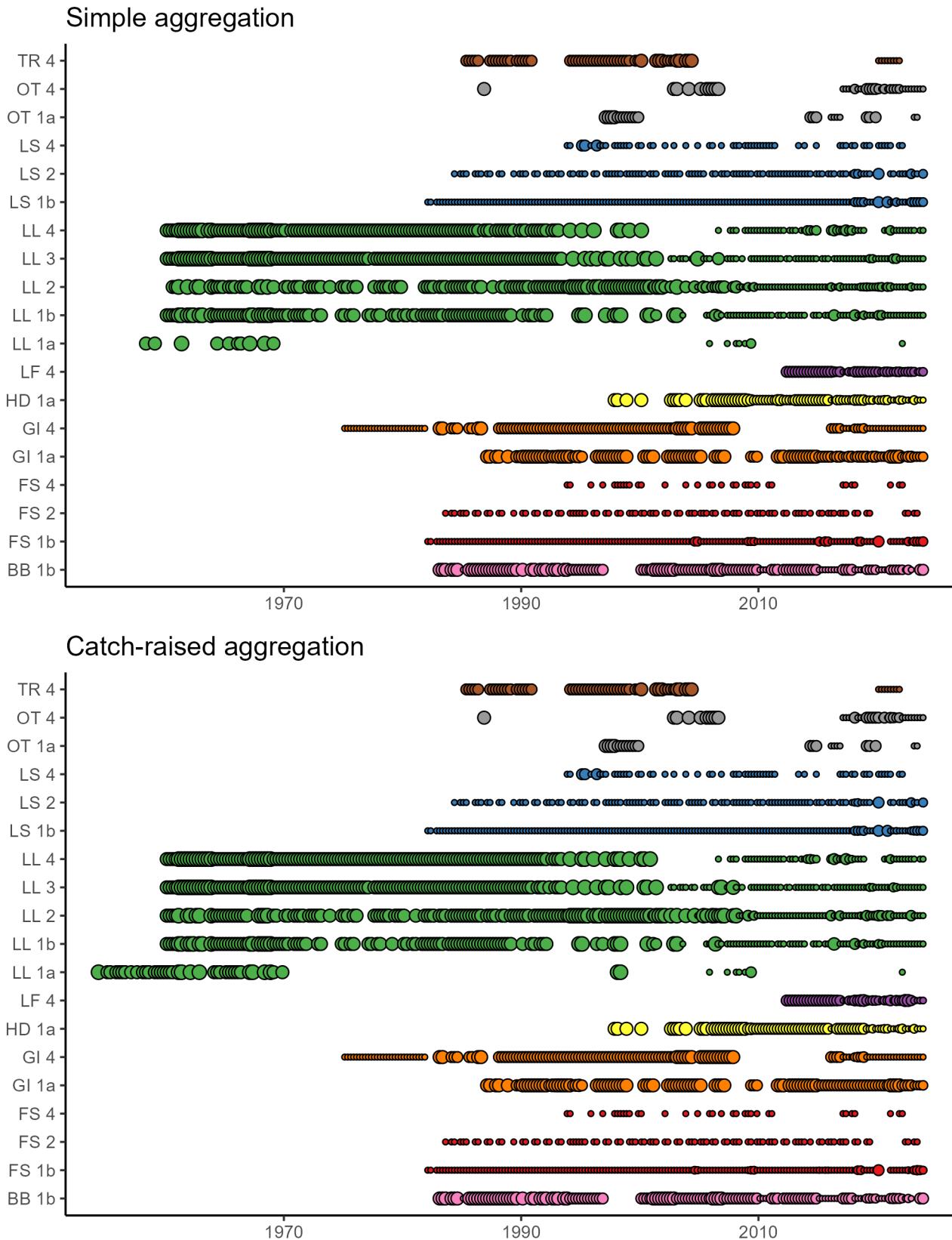


Figure 7: The availability of size composition data from each fishery by quarter. The size of the bubble indicates the reporting quality score (small size = high quality, large size = low quality).

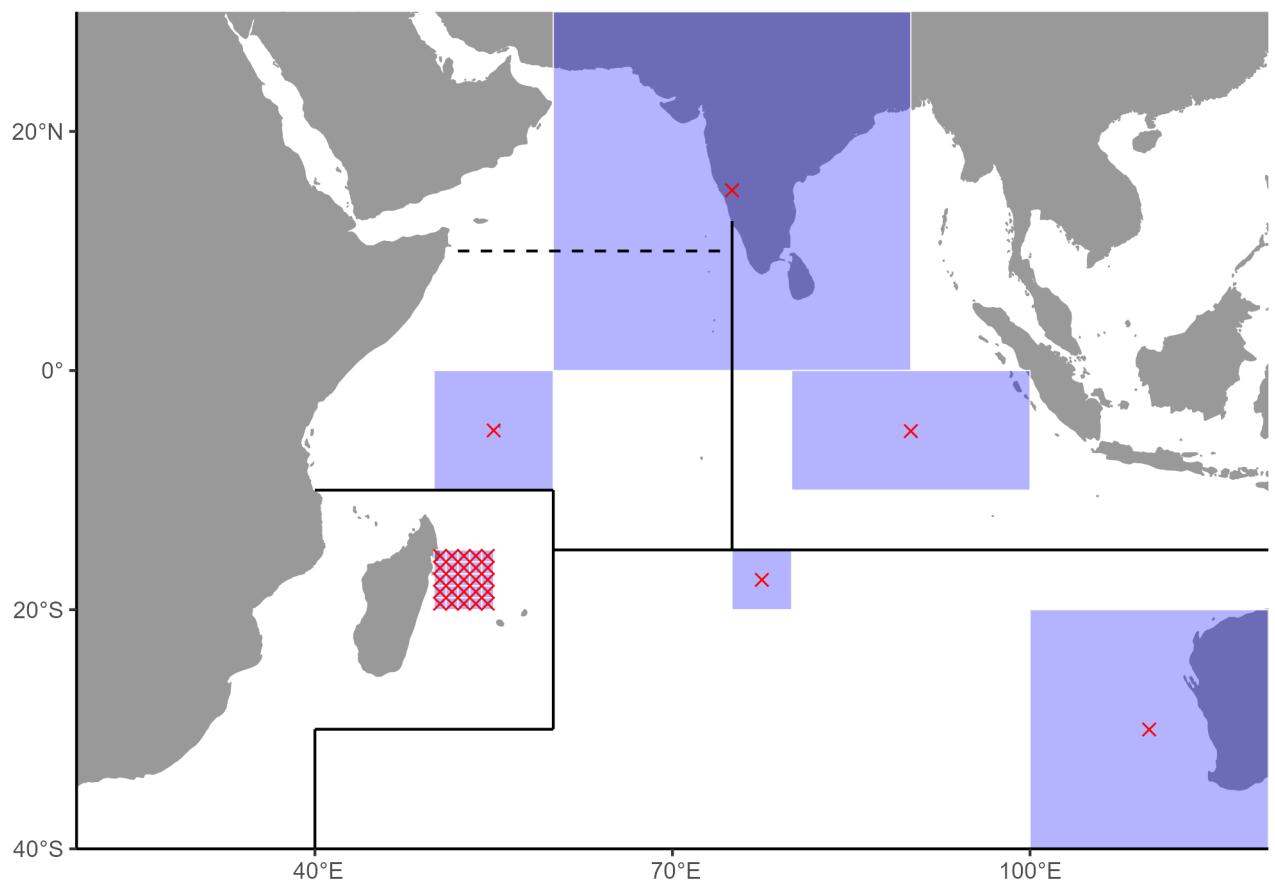


Figure 8: Example of grid categories in the original size dataset. The region assignment was done based on the grid centroid.

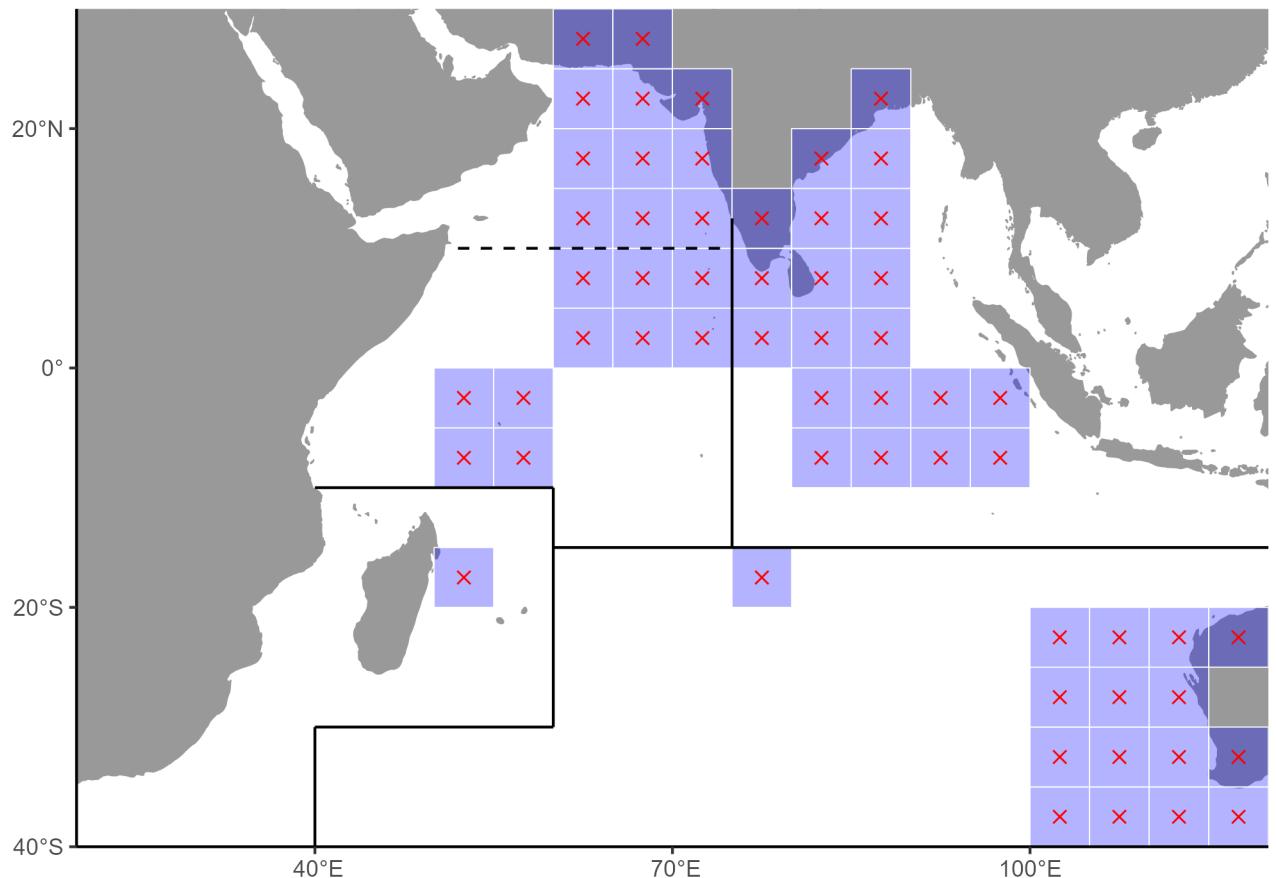


Figure 9: Example of grids in the cwp55 size dataset. The region assignment was done based on the grid centroid.

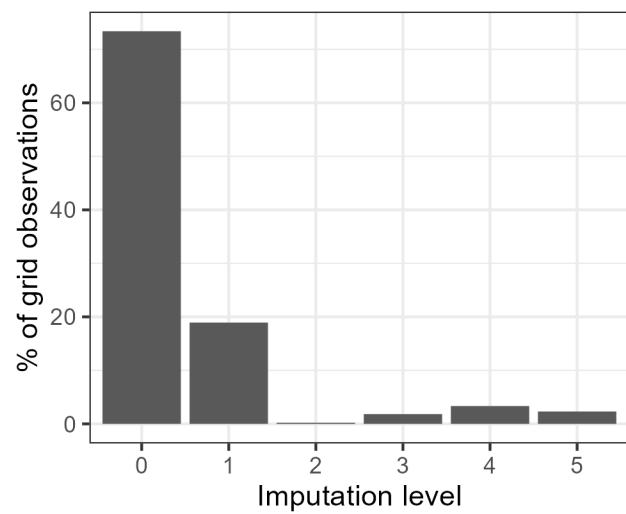


Figure 10: Percentage of size observations by imputation level. Level 0 means perfect match, so no imputation was required.

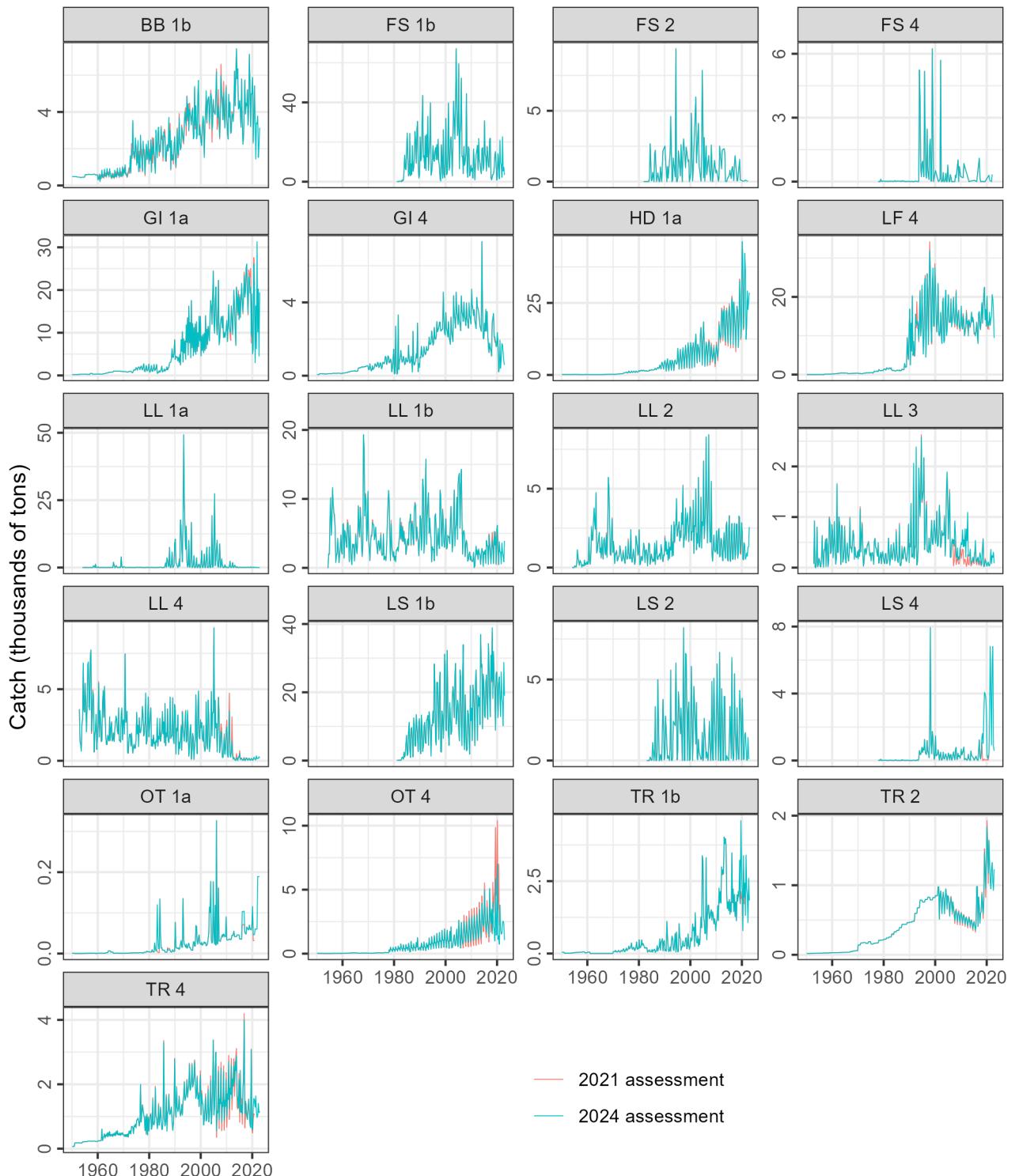


Figure 11: Comparison between the catch values used in the 2021 assessment and the current (2024) values.

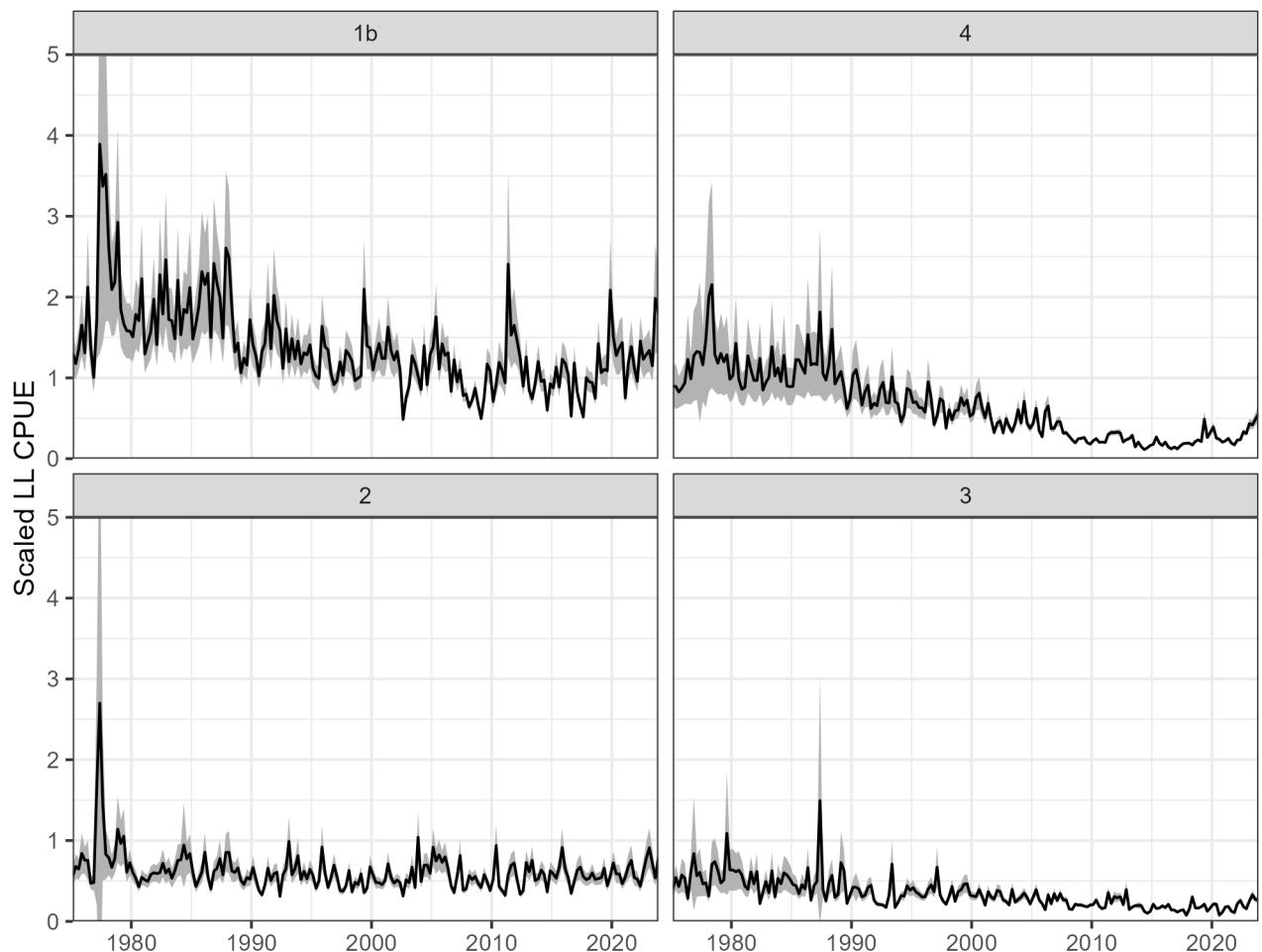


Figure 12: Scaled CPUE LL time serie per region. The shaded area is the 95% confidence interval.

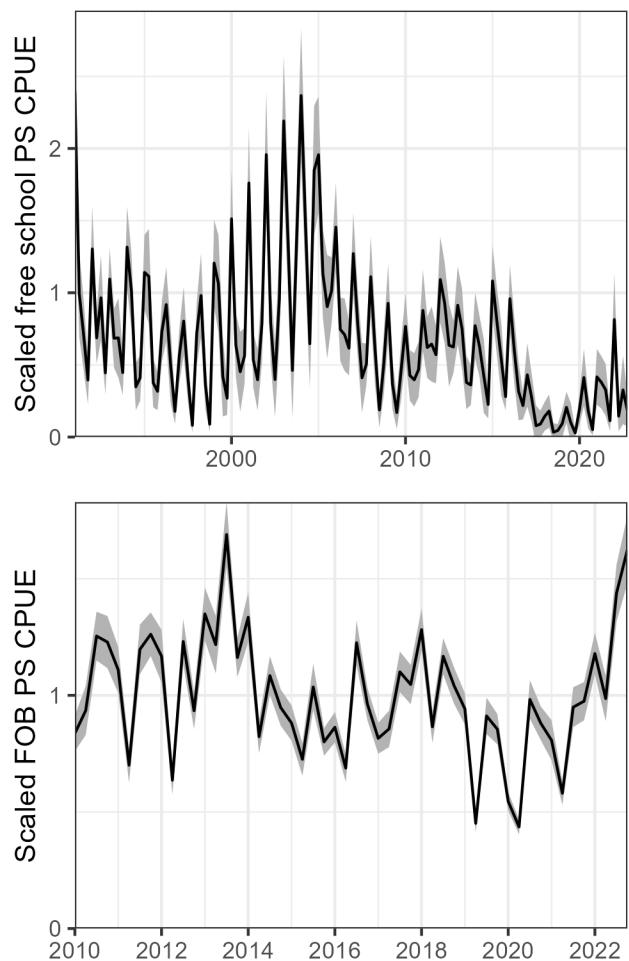


Figure 13: Scaled purse seine CPUE time series. Both series were used for region 1b. The shaded area is the 95% confidence interval.

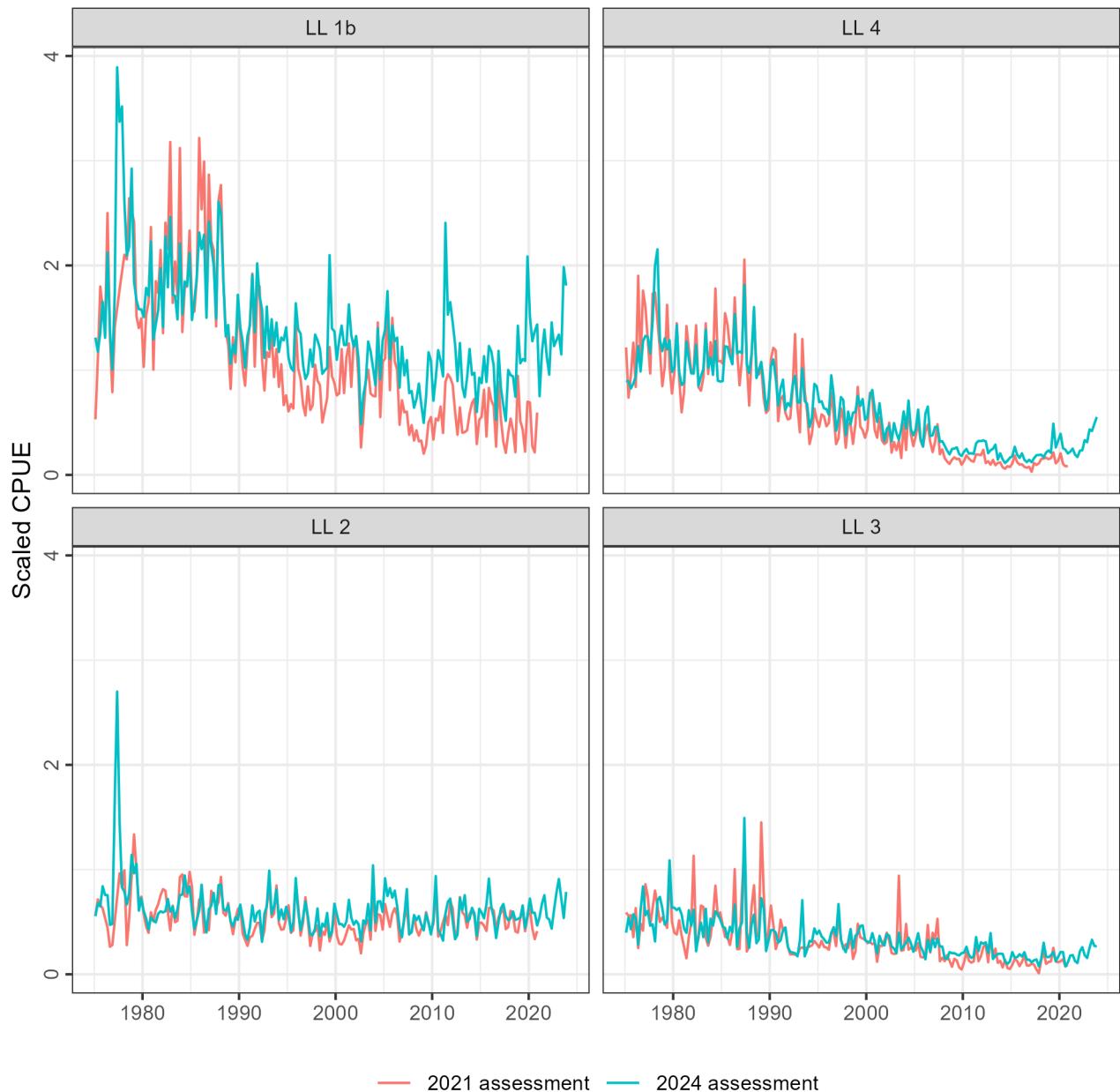


Figure 14: Comparison between the scaled CPUE LL values used in the 2021 assessment and the current (2024) values.

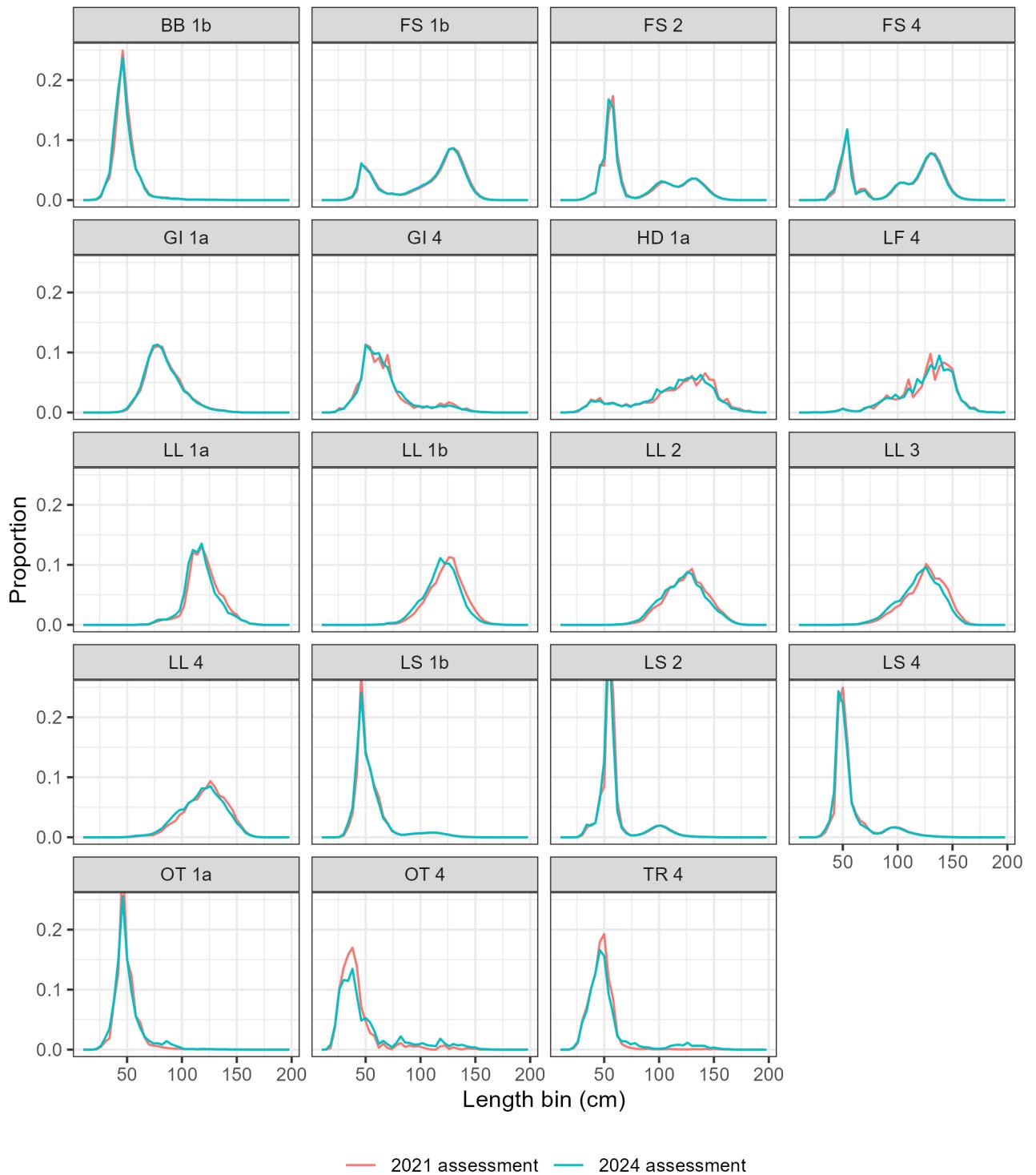


Figure 15: Comparison between the aggregated size compositions per fishery used in the 2021 and the current (2024) assessment.

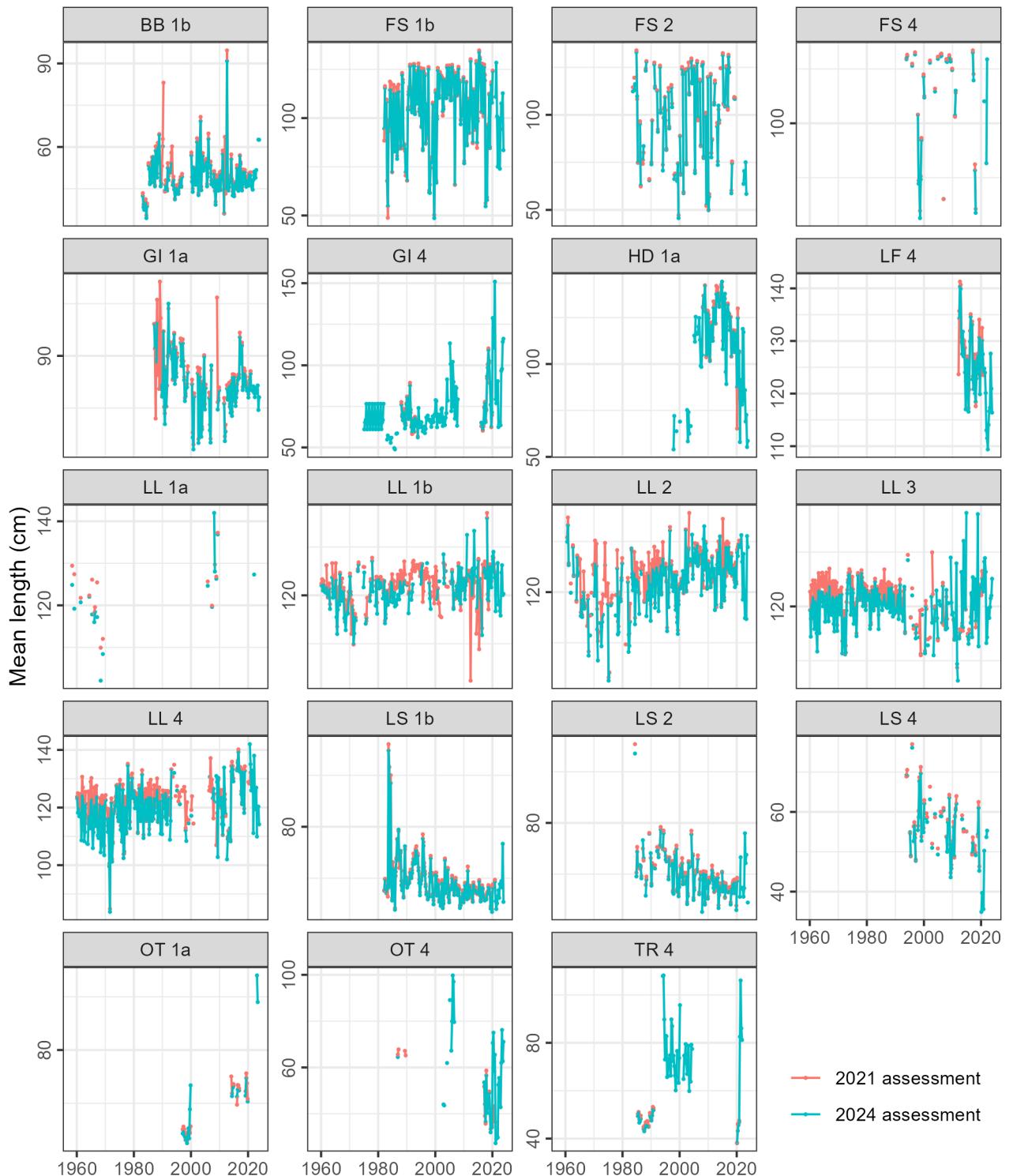


Figure 16: Comparison between the mean length per quarter per fishery used in the 2021 and in the current (2024) assessment.

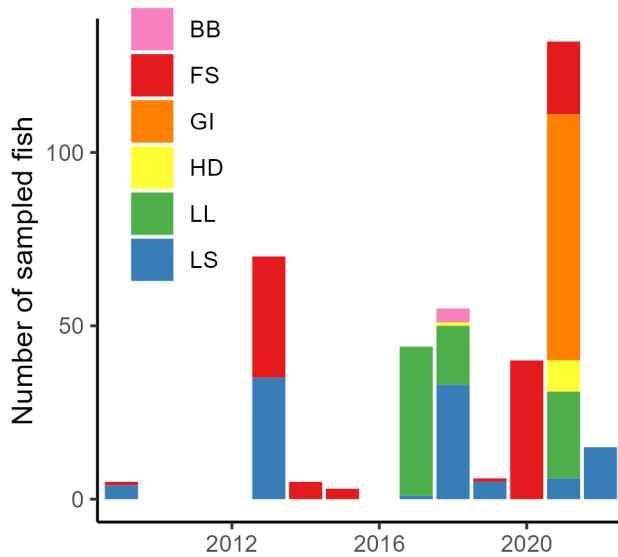


Figure 17: Number of sampled fish per year and fishery group in the age-length dataset from the GERUNDIO project.

12.2 Clustering of size compositions

Add text here.

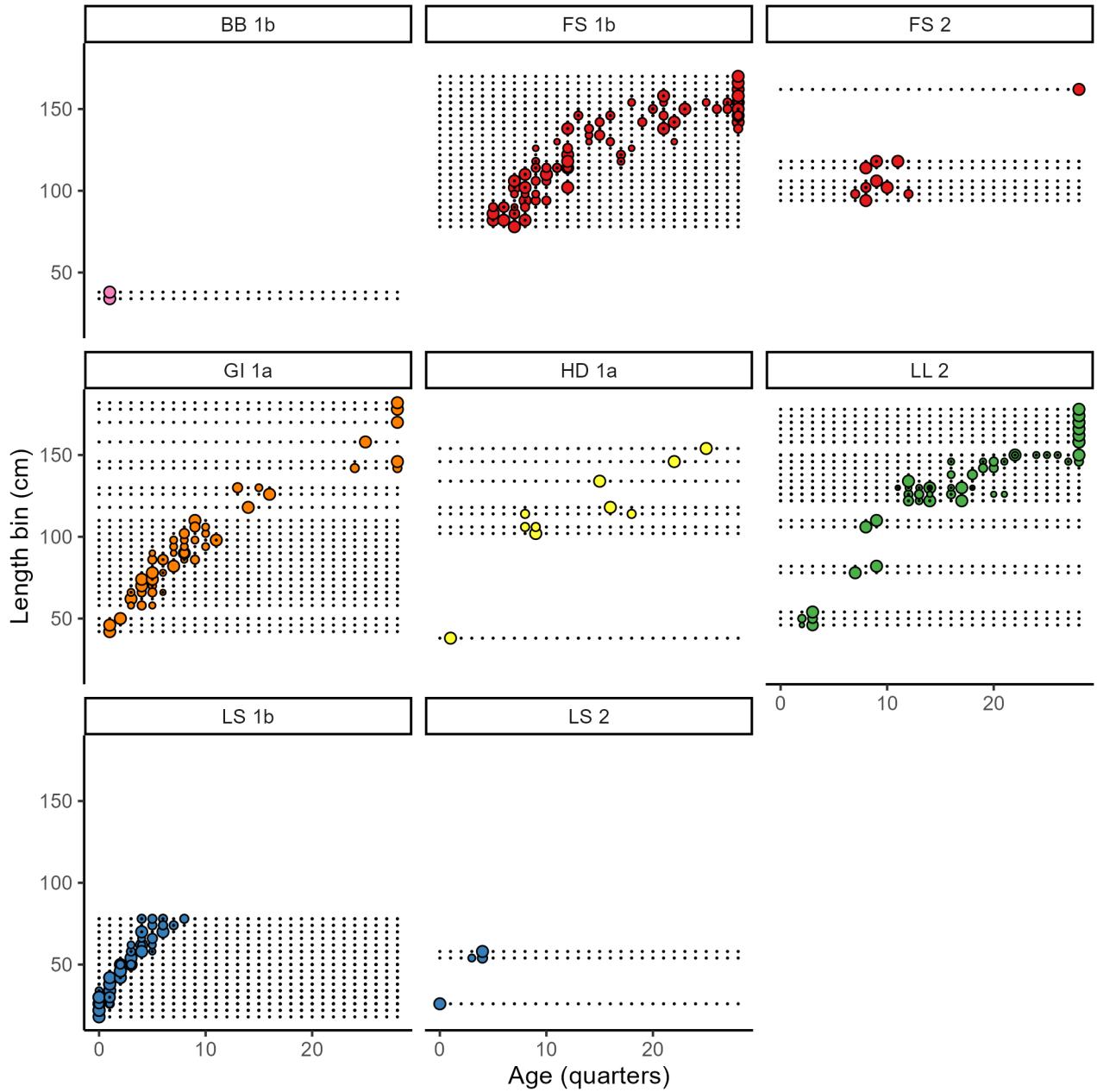


Figure 18: Conditional age-at-length (CAAL) data included in the assessment model. The bubble size represents the proportion of ages at a given length.

12.3 One-area model configuration

Add text here.

12.4 Two-areas model configuration

Add text here.

References

- Artetxe-Arrate, Iraide, Igaratza Fraile, Patricia Lastra-Luque, Jessica Farley, Naomi Clear, Umair Shahid, S. Abdul-Razzaue, et al. in review. "Otolith Stable Isotopes Highlight the Importance of Local Nursery Areas as the Origin of Recruits to Yellowfin Tuna (*Thunnus Albacares*) Fisheries in the Western Indian Ocean." *Fisheries Research*, in review.
- Artetxe-Arrate, Iraide, Igaratza Fraile, Francis Marsac, Jessica H. Farley, Naiara Rodriguez-Ezpeleta, Campbell R. Davies, Naomi P. Clear, Peter Grewe, and Hilario Murua. 2021. "A Review of the Fisheries, Life History and Stock Structure of Tropical Tuna (Skipjack Katsuwonus Pelamis, Yellowfin *Thunnus Albacares* and Bigeye *Thunnus Obesus*) in the Indian Ocean." In *Advances in Marine Biology*, 88:39–89. Elsevier. <https://doi.org/10.1016/bs.amb.2020.09.002>.
- Chassot, Emmanuel. 2014. "Are There Small Yellowfin Caught by Purse Seiners in Free-Swimming Schools?" IOTC-2014-WPDCS10-INF05. Indian Ocean Tuna Comission.
- Correa, Giancarlo M., Jon Uranga, David Kaplan, Gorka Merino, and Lourdes Ramos. 2024. "Standardized Catch Per Unit Effort of Yellowfin Tuna in the Indian Ocean for the European Purse Seine Fleet Operating on Floating Objects." IOTC-2024-WPTT26(DP)-11rev1. Indian Ocean Tuna Comission.
- Dammannagoda, Sudath T., David A. Hurwood, and Peter B. Mather. 2008. "Evidence for Fine Geographical Scale Heterogeneity in Gene Frequencies in Yellowfin Tuna (*Thunnus Albacares*) from the North Indian Ocean Around Sri Lanka." *Fisheries Research* 90 (1-3): 147–57. <https://doi.org/10.1016/j.fishres.2007.10.006>.
- Duffy, Leanne M., Petra M. Kuhnert, Heidi R. Pethybridge, Jock W. Young, Robert J. Olson, John M. Logan, Nicolas Goñi, et al. 2017. "Global Trophic Ecology of Yellowfin, Bigeye, and Albacore Tunas: Understanding Predation on Micronekton Communities at Ocean-Basin Scales." *Deep Sea Research Part II: Topical Studies in Oceanography* 140 (June): 55–73. <https://doi.org/10.1016/j.dsr2.2017.03.003>.
- Farley, Jessica H., Kyne KrusicGolub, Paige Eveson, Patricia Luque, Igaratza Fraile, Iraide Artetxe-Arrate, Iker Zudaire, et al. 2023. "Updating the Estimation of Age and Growth of Yellowfin Tuna (*Thunnus Albacares*) in the Indian Ocean Using Otoliths." IOTC-2023-WPTT25-20. Indian Ocean Tuna Comission.
- Farley, Jessica, Kyne Krusic-Golub, Paige Eveson, Patricia Luque, Naomi Clear, Igaratza Fraile, Iraide Artetxe-Arrate, et al. 2021. "Estimating the Age and Growth of Yellowfin Tuna (*Thunnus Albacares*) in the Indian Ocean from Counts of Daily and Annual Increments in Otoliths." IOTC-2021-WPTT23-05_Rev1. Indian Ocean Tuna Comission.
- Farley, Jessica, Krusic-Golub, Paige Eveson, Naomi Clear, F. Rouspard, C. Sanchez, Simon Nicol, and John Hampton. 2020. "Age and Growth of Yellowfin and Bigeye Tuna in the Western and Central Pacific Ocean from Otoliths." WCPFC-SC16-2020/SC16-SA-WP-02. Western and Central Pacific Fisheries Commission.
- Froese, Rainer, and Daniel Pauly. 2024. "FishBase." World {{Wide Web}} Electronic Publication. www.fishbase.org.
- Fu, Dan. 2017. "Indian Ocean Skipjack Tuna Stock Assessment 1950-2016 (Stock Synthesis)." IOTC-2017-WPTT19-47. Indian Ocean Tuna Comission.
- . 2020. "Tag Data Processing for IOTC Tropical Tuna Assessments." IOTC-2020-WPTT22(DP)-10. Indian Ocean Tuna Comission.
- Fu, Dan, Adam D. Langley, Gorka Merino, and Agurtzane Urtizberea. 2018. "Preliminary Indian Ocean Yellowfin Tuna Stock Assessment 1950-2017 (Stock Synthesis)." IOTC-2018-WPTT20-33. Indian Ocean Tuna Comission.
- Fu, Dan, Agurtzane Urtizberea Ijurco, Massimiliano Cardinale, Richard D Methot, Simon D.

- Hoyle, and Gorka Merino. 2021. "Preliminary Indian Yellowfin Tuna Stock Assessment 1950-2020 (Stock Synthesis)." IOTC-2021-WPTT23-12. Indian Ocean Tuna Comission.
- Gaertner, Daniel, and Jean Pierre Hallier. 2015. "Tag Shedding by Tropical Tunas in the Indian Ocean and Other Factors Affecting the Shedding Rate." *Fisheries Research* 163 (March): 98–105. <https://doi.org/10.1016/j.fishres.2014.02.025>.
- Greehan, James, and Simon D. Hoyle. 2013. "Review of Length Frequency Data of the Taiwanese Distant Water Longline Fleet." IOTC-2013-WPDCS09-12. Indian Ocean Tuna Comission.
- Grewe, Peter, Pierre Feutry, Scott Foster, Aulich, Matt Lansdell, Scott Cooper, Naomi Clear, et al. 2020. "Genetic Population Connectivity of Yellowfin Tuna in the Indian Ocean from the PSTBS-IO Project." IOTC-2020-WPTT22(AS)12_REV1. Seychelles: Indian Ocean Tuna Comission.
- Herrera, Miguel. 2010. "Proposal for a System to Assess the Quality of Fisheries Statistics at the IOTC." IOTC-2010-WPDCS-06. Indian Ocean Tuna Comission.
- Hillary, R., Secretariat IOTC, and J. Areso. 2008. "Reporting Rate Analyses for Recaptures from Seychelles Port for Yellowfin, Bigeye and Skipjack Tuna." IOTC-2008-WPTT-18. Indian Ocean Tuna Comission.
- Hillary, R., J. Million, A. Anganuzzi, and J. Areso. 2008. "Tag Shedding and Reporting Rate Estimates for Indian Ocean Tuna Using Double-Tagging and Tag-Seeding Experiments." IOTC-2008-WPTDA-04. Indian Ocean Tuna Comission.
- Hosseini, S A, and F Kaymaram. 2016. "Investigations on the Reproductive Biology and Diet of Yellowfin Tuna, *Thunnus Albacares*, (Bonnaterre, 1788) in the Oman Sea." *Journal of Applied Ichthyology* 32: 310–17. <https://doi.org/10.1111/jai.12907>.
- Hoyle, Simon D. 2018. "Indian Ocean Tropical Tuna Regional Scaling Factors That Allow for Seasonality and Cell Areas." IOTC-2018-WPM09-13. Indian Ocean Tuna Comission.
- . 2021. "Review of Size Data from Indian Ocean Longline Fleets, and Its Utility for Stock Assessment." IOTC-2021-WPTT23-07. Indian Ocean Tuna Comission.
- . 2024. "Effort Creep in Longline and Purse Seine CPUE and Its Application in Tropical Tuna Assessments." IOTC-2024-WPTT26(DP)-16. Indian Ocean Tuna Comission.
- Hoyle, Simon D., Robert A. Campbell, Nicholas D. Ducharme-Barth, Arnaud Grüss, Bradley R. Moore, James T. Thorson, Laura Tremblay-Boyer, Henning Winker, Shijie Zhou, and Mark N. Maunder. 2024. "Catch Per Unit Effort Modelling for Stock Assessment: A Summary of Good Practices." *Fisheries Research* 269 (January): 106860. <https://doi.org/10.1016/j.fishres.2023.106860>.
- Hoyle, Simon D., Bruno M. Leroy, Simon J. Nicol, and W. John Hampton. 2015. "Covariates of Release Mortality and Tag Loss in Large-Scale Tuna Tagging Experiments." *Fisheries Research* 163 (March): 106–18. <https://doi.org/10.1016/j.fishres.2014.02.023>.
- Hoyle, Simon D., Keisuke Satoh, and Takayuki Matsumoto. 2017. "Exploring Possible Causes of Historical Discontinuities in Japanese Longline CPUE." IOTC-2017-WPTT19-33. Indian Ocean Tuna Comission.
- IOTC, Secretariat. 2021. "Review of Yellowfin Tuna Statistical Data." IOTC-2021-WPTT23(DP)-07_Rev1. Indian Ocean Tuna Comission.
- . 2024. "Review of the Statistical Data Available for Yellowfin Tuna (1950-2022)." IOTC-2024-WPTT26(DP)-07. Indian Ocean Tuna Comission.
- Kaplan, David, Giancarlo M Correa, Lourdes Ramos, Antoine Duparc, Jon Uranga, Josu Santiago, Laurent Floch, et al. 2024. "Standardized CPUE Abundance Indices for Adult Yellowfin Tuna Caught in Free-Swimming School Sets by the European Purse-Seine Fleet in the Indian Ocean, 1991-2022." IOTC-2024-WPTT26(DP)-13rev2.
- Kolody, D. 2018. "Estimation of Indian Ocean Skipjack Purse Seine Catchability Trends from

- Bigeye and Yellowfin Assessments.” IOTC–2018–WPTT20–32. Indian Ocean Tuna Comission.
- Kolody, D., Miguel Herrera, and Julien Million. 2011. “Indian Ocean Skipjack Tuna Stock Assessment 1950–2009 (Stock Synthesis).” IOTC-2011-WPTT13-31. Indian Ocean Tuna Comission.
- Krishnan, Silambarasan, Tiburtius Antony Pillai, John Chembian Antony Rayappan, Tharumar Yagappan, and Jeyabaskaran Rajapandian. 2024. “Diet Composition and Feeding Habits of Yellowfin Tuna *Thunnus Albacares* (Bonnaterre, 1788) from the Bay of Bengal.” *Aquatic Living Resources* 37: 10. <https://doi.org/10.1051/alr/2024008>.
- Kumar, Mamidi Satish, and Shubhadeep Ghosh. 2022. “Reproductive Dynamics of Yellowfin Tuna, *Thunnus Albacares* (Bonnaterre 1788) Exploited from Western Bay of Bengal.” *Thalassas: An International Journal of Marine Sciences* 38: 1003–12. <https://doi.org/10.1007/s41208-022-00429-1>.
- Kunal, Swaraj Priyaranjan, Girish Kumar, Maria Rosalia Menezes, and Ram Murti Meena. 2013. “Mitochondrial DNA Analysis Reveals Three Stocks of Yellowfin Tuna *Thunnus Albacares* (Bonnaterre, 1788) in Indian Waters.” *Conservation Genetics* 14 (1): 205–13. <https://doi.org/10.1007/s10592-013-0445-3>.
- Lan, Kuo-Wei, Yi-Jay Chang, and Yan-Lun Wu. 2020. “Influence of Oceanographic and Climatic Variability on the Catch Rate of Yellowfin Tuna (*Thunnus Albacares*) Cohorts in the Indian Ocean.” *Deep Sea Research Part II: Topical Studies in Oceanography* 175 (May): 104681. <https://doi.org/10.1016/j.dsro.2019.104681>.
- Lan, Kuo-Wei, Karen Evans, and Ming-An Lee. 2013. “Effects of Climate Variability on the Distribution and Fishing Conditions of Yellowfin Tuna (*Thunnus Albacares*) in the Western Indian Ocean.” *Climatic Change* 119 (1): 63–77. <https://doi.org/10.1007/s10584-012-0637-8>.
- Langley, Adam D. 2015. “Stock Assessment of Yellowfin Tuna in the Indian Ocean Using Stock Synthesis.” IOTC–2015–WPTT17–30. Indian Ocean Tuna Comission.
- . 2016. “An Update of the 2015 Indian Ocean Yellowfin Tuna Stock Assessment for 2016.” IOTC-2016-WPTT18-27. Indian Ocean Tuna Comission.
- Langley, Adam D., Dan Fu, and Mark Maunder. 2023. “An Investigation of the Recruitment Dynamics of Indian Ocean Yellowfin Tuna.” IOTC-2023-WPTT25-12. Indian Ocean Tuna Comission.
- Langley, Adam D., John Hampton, Miguel Herrera, and Julien Million. 2008. “Preliminary Stock Assessment of Yellowfin Tuna in the Indian Ocean Using MULTIFAN-CL.” IOTC-2008-WPTT-10. Indian Ocean Tuna Comission.
- Langley, Adam D., Miguel Herrera, Jean-Pierre Hallier, and Julien Million. 2009. “Stock Assessment of Yellowfin Tuna in the Indian Ocean Using MULTIFAN-CL.” IOTC-2009-WPTT-10. Indian Ocean Tuna Comission.
- Langley, Adam D., Miguel Herrera, and Julien Million. 2010. “Stock Assessment of Yellowfin Tuna in the Indian Ocean Using MULTIFAN-CL.” IOTC-2010-WPTT-23. Indian Ocean Tuna Comission.
- . 2011. “Stock Assessment of Yellowfin Tuna in the Indian Ocean Using MULTIFAN-CL.” IOTC-2011-WPTT-13. Indian Ocean Tuna Comission.
- . 2012. “Stock Assessment of Yellowfin Tuna in the Indian Ocean Using MULTIFAN-CL.” IOTC-2012-WPTT-14-38 Rev_1. Indian Ocean Tuna Comission.
- Lee, Huihua, Kevin R. Piner, Ian G. Taylor, and Toshihide Kitakado. 2019. “On the Use of Conditional Age at Length Data as a Likelihood Component in Integrated Population Dynamics Models.” *Fisheries Research* 216 (August): 204–11. <https://doi.org/10.1016/j.fishres.2019.04.007>.

- Matsumoto, Takayuki, Keisuke Satoh, Wen-Pei Tsai, Sheng-Ping Wang, Jung-Hyun Lim, Heewon Park, and Sung Il Lee. 2024. "Joint Longline CPUE for Yellowfin Tuna in the Indian Ocean by the Japanese, Korean and Taiwanese Longline Fishery." IOTC-2024-WPTT26(DP)-14. Indian Ocean Tuna Comission.
- Maunder, Mark, Carolina V. Minte-Vera, Adam D. Langley, and Daniel Howell. 2023. "Independent Review of Recent IOTC Yellowfin Tuna Assessment." IOTC-2023-WPTT25-13_Rev1. Indian Ocean Tuna Comission.
- McKechnie, S., G. Pilling, and John Hampton. 2017. "Stock Assessment of Bigeye Tuna in the Western and Central Pacific Ocean." WCPFC-SC13-2017/SA-WP-05. Western and Central Pacific Fisheries Commission.
- Ménard, Frédéric, Anne Lorrain, Michel Potier, and Francis Marsac. 2007. "Isotopic Evidence of Distinct Feeding Ecologies and Movement Patterns in Two Migratory Predators (Yellowfin Tuna and Swordfish) of the Western Indian Ocean." *Marine Biology* 153 (2): 141–52. <https://doi.org/10.1007/s00227-007-0789-7>.
- Methot, Richard D. 2019. "Recommendations on the Configuration of the Indian Ocean Yellowfin Tuna Stock Assessment Model."
- Methot, Richard D., and Chantell R. Wetzel. 2013. "Stock Synthesis: A Biological and Statistical Framework for Fish Stock Assessment and Fishery Management." *Fisheries Research* 142 (May): 86–99. <https://doi.org/10.1016/j.fishres.2012.10.012>.
- Moore, Bradley R, Pratiwi Lestari, Scott C Cutmore, Craig Proctor, and Robert J G Lester. 2019. "Movement of Juvenile Tuna Deduced from Parasite Data." Edited by James Watson. *ICES Journal of Marine Science* 76 (6): 1678–89. <https://doi.org/10.1093/icesjms/fsz022>.
- Muhling, Barbara A., John T. Lamkin, Francisco Alemany, Alberto García, Jessica Farley, G. Walter Ingram, Diego Alvarez Berastegui, Patricia Reglero, and Raul Laiz Carrion. 2017. "Reproduction and Larval Biology in Tunas, and the Importance of Restricted Area Spawning Grounds." *Reviews in Fish Biology and Fisheries* 27 (4): 697–732. <https://doi.org/10.1007/s11160-017-9471-4>.
- Nishida, Tom, and Hiroshi Shono. 2005. "Stock Assessment of Yellowfin Tuna (*Thunnus Albacares*) Resources in the Indian Ocean by the Age Structured Production Model (ASPM) Analyses." IOTC-2005-WPTT-09. Indian Ocean Tuna Comission.
- . 2007. "Stock Assessment of Yellowfin Tuna (*Thunnus Albacares*) in the Indian Ocean by the Age Structured Production Model(ASPM) Analyses." IOTC-2007-WPTT-12. Indian Ocean Tuna Comission.
- Nootmorn, Praulai, Anchalee Yakoh, and Kannokwan Kawises. 2005. "Reproductive Biology of Yellowfin Tuna in the Eastern Indian Ocean." IOTC-2005-WPTT-14. Indian Ocean Tuna Comission.
- Pecoraro, C., I. Zudaire, N. Bodin, H. Murua, P. Taconet, P. Díaz-Jaimes, A. Cariani, F. Tinti, and E. Chassot. 2017. "Putting All the Pieces Together: Integrating Current Knowledge of the Biology, Ecology, Fisheries Status, Stock Structure and Management of Yellowfin Tuna (*Thunnus Albacares*)." *Reviews in Fish Biology and Fisheries* 27 (4): 811–41. <https://doi.org/10.1007/s11160-016-9460-z>.
- Punt, André E. 2019. "Spatial Stock Assessment Methods: A Viewpoint on Current Issues and Assumptions." *Fisheries Research* 213 (May): 132–43. <https://doi.org/10.1016/j.fishres.2019.01.014>.
- Roger, Claude. 1994. "Relationships Among Yellowfin and Skipjack Tuna, Their Prey-Fish and Plankton in the Tropical Western Indian Ocean." *Fisheries Oceanography* 3 (2): 133–41. <https://doi.org/10.1111/j.1365-2419.1994.tb00055.x>.
- Sabarros, Philippe, Evgeny Romanov, and Pascal Bach. 2015. "Vertical Behavior and Habitat Preferences of Yellowfin and Bigeye Tuna in the South West Indian Ocean Inferred from

- PSAT Tagging Data.” IOTC-2015-WPTT17-42 Rev_1. Indian Ocean Tuna Comission.
- Urtizberea, Agurtzane, Massimiliano Cardinale, Henning Winker, Richard D. Methot, Dan Fu, Toshihide Kitakado, Carmen Fernandez, and Gorka Merino. 2020. “Towards Providing Scientific Advice for Indian Ocean Yellowfin in 2020.” IOTC-2020-WPTT22(AS)-21. Indian Ocean Tuna Comission.
- Urtizberea, Agurtzane, Dan Fu, Gorka Merino, Richard D Methot, Massimiliano Cardinale, Henning Winker, John Walter, and Hilario Murua. 2019. “Preliminary Assessment of Indian Ocean Yellowfin Tuna 1950-2018 (Stock Synthesis, V3.30).” IOTC-2019-WPTT21-50. Indian Ocean Tuna Comission.
- Wain, Gwenaëlle, Lorelei Guéry, David Michael Kaplan, and Daniel Gaertner. 2021. “Quantifying the Increase in Fishing Efficiency Due to the Use of Drifting FADs Equipped with Echosounders in Tropical Tuna Purse Seine Fisheries.” Edited by Richard O’Driscoll. *ICES Journal of Marine Science* 78 (1): 235–45. <https://doi.org/10.1093/icesjms/fsaa216>.
- Zudaire, Iker, Iraide Artetxe-Arrate, Jessica H. Farley, Hilario Murua, Deniz Kukul, Annie Vidot, Shoaib Razzaque, et al. 2022. “Preliminary Estimates of Sex Ratio, Spawning Season, Batch Fecundity and Length at Maturity for Indian Ocean Yellowfin Tuna.” IOTC-2022-WPTT24(DP)-09. Indian Ocean Tuna Comission.
- Zudaire, Iker, Hilario Murua, Maitane Grande, and Nathalie Bodin. 2013. “Reproductive Potential of Yellowfin Tuna (*Thunnus Albacares*) in the Western Indian Ocean.” *Fishery Bulletin* 111 (3): 252–64. <https://doi.org/10.7755/FB.111.3.4>.