

Article

Study Of Albacore Tuna (*Thunnus alalunga*) Abundance Using Regional Ocean Modeling System (ROMS) Data In Indian Ocean

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Abstract

First, we analysed alabcore catch data based on time, positions, and layer alabcore caught and ROMS result data monthly climatology data for temperature, salinity. current velocity, and sea surface height for 2005–2008. Then, we analyzed the relationship between catch data and ROMS data by combining the statistical method of regression trogh origin (RTO) and geographic information system (GIS). Three model RTO were generated with the abundance of albacore tuna as a response variable, and temperature, salinity. current velocity, and sea surface height as predictor variables. All of the predictors of temperature, salinity. current velocity, and sea surface height were highly significant ($P < 0.001$) to the number of albacore tuna. Values of temperature, salinity. current velocity, and sea surface height in albacore tuna preferences ranged from 220 to 230 C, 34.79 to 34.84 Psu, 0.01 to 0.03 m/s and 0.66 to 0.70 m, respectively. Validation of the predicted number of albacore tuna with the observed value was significant ($P < 0.05$, $r^2 = 0.60$). sea surface height was the most important environmental variable to the number of albacore tuna caught, followed by temperature, salinity and current velocity.

Keywords: albacore; abundance; ROMS; Indian Ocean

1. Introduction

Albacore (*Thunnus alalunga*; hereafter ALB) is one of the main target species of the commercial tuna fishery and has a long history of scientific research (Chen et al., 2005). Although albacore tuna has a long history of scientific research and remains the preeminent target species of the commercial tuna fishery, the related literature in the Indian Ocean is scarce in comparison with other oceanic basins like the Atlantic and Pacific Oceans (Nikolic et al., 2014). The geographical distribution of ALB is from 45–50°N to 30–40°S, although it is less abundant in surface waters between 10°N and 10°S (Collette & Nauen, 1983). In the Indian Ocean, ALB is distributed mainly in areas south of 10°S and it migrates seasonally between north and south areas in the Southern Hemisphere (Chen et al., 2005). In the Indian Ocean, ALB is caught almost exclusively under drifting longline (98%), with remaining catches recorded under purse seines and other gears (IOTC, 2007; Nishida & Tanaka, 2008).

Cacth of ALB by Indonesian longline fleets operating in Indian Ocean from 2004-2006 was estimated at 9,081 tons by IOTC, while 53.4% of which was landed at Bena fishing port (Proctor et al., 2007). Although the Indian Ocean is one of the most productive fishing grounds, and has a long history of tuna exploitation, little is known about the oceanographic conditions affecting the distribution and abundance of ALB. Previous

studies showed that Temperature is the most important parameter ALB abundance in the Indian Ocean (Lan et al., 2011; Chen et al., 2005). Whereas for other oceanographic parameters have not been too many studied in the Indian Ocean compared to research in the Pacific Ocean. Goñi et al. (2015) used ROMS to investigate the influence of temperature, salinity and trophic parameters on albacore daily catches in the Northeast Atlantic. They found that the thermal stratification seems to have an important effect on Albacore catches.

The habitat is highly influenced by dynamic oceanographic factors, different characteristic of oceanographic variabilities can influence on fish distribution. Changes and variations of oceanographic factors indicate that the abundance of fish in the waters is unlevel (Zainuddin, 2007). Potential fishing grounds are always changing and moving in accordance with the environmental factors. Fish will choose a more suitable habitat for feeding, shelter, reproduction and migration (Palacios et al., 2006). Generally, the fishing activities by the fisherman are only based on repeated experiences as well as gathering information from fellow fishermen. They are often in uncertain about location of the potential fishing grounds. This leads to wastage of money and time, and catches are less than optimal which impact on the income of fishermen.

The first aim of this research is to find out relationship between oceanographic variables from Regional Ocean Modeling System (ROMS) data on albacore tuna abundance in Indian Ocean. ROMS incorporates advanced features and high order numerics, allowing efficient and robust resolution of mesoscale dynamics in the oceanic and coastal domains (ROMS; Shchepetkin and McWilliams, 2005). The model solves the free surface, hydrostatic, primitive equations of the fluid dynamics over variable topography using stretched, terrain-following coordinates in the vertical and orthogonal, curvilinear coordinates in the horizontal (Putra et al., 2009). ROMS have a high accuracy rate that is suitable for investigating vertically Oceanographic conditions so expect to get maximum results in this study.

2. Materials and Methods

2.1 Study Area

The study location is the range between 09°-15° S latitude and 110°-120° E Longitude. The area was selected based on data availability and information ALB fishing operations (Figure 1).

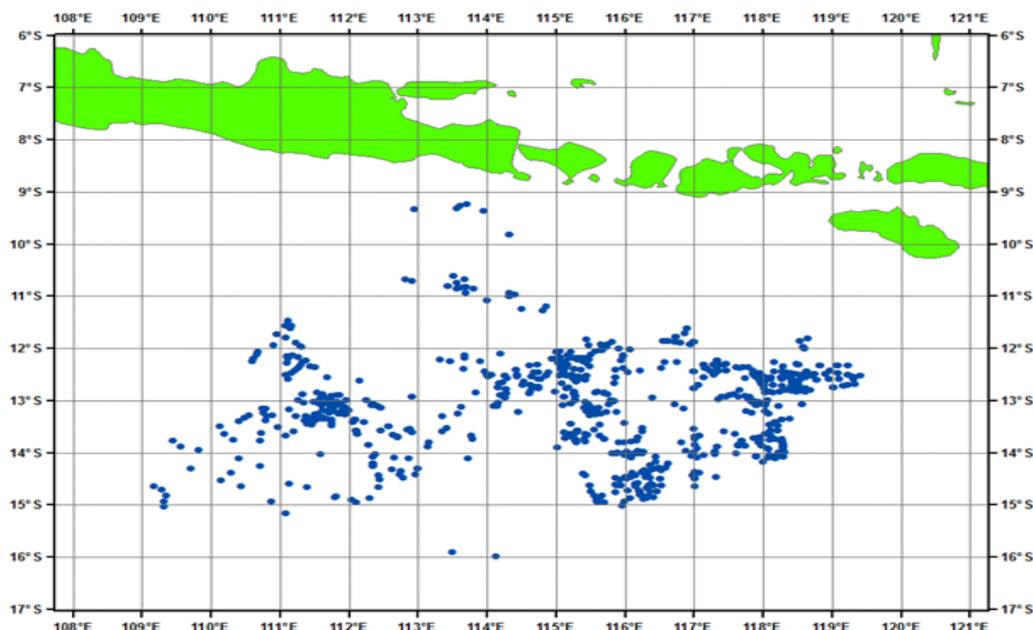


Figure 1. Research location

2.2 Analysis Data

Various variables were analyzed on this research, includes ALB catchment data, in addition oceanographic variables consist of temperature, salinity, current velocity are results from ROMS data (1990-2008) is divided into 6 layers except for the sea surface heigth which were analyzed to examine their influences on ALB tuna data.

2.2.1. Oceanographic Data

ROMS data divided into two, the data consist forcing data and boundary condition data. Forcing data using NCEP reanalysis data and boundary condition data use SODA data (Simple Ocean Data Assimilation). ROMS data Spatial resolution is 10km for the horizontal and 25 S-coordinate layer vertically. Initial Condition using SODA with updates boundary of each month the data U, V, T, and S, with models running from 1990 to 2008. The model domain of $90^{\circ} - 150^{\circ}$ E and 15° S - 15° N, but in the domain visualization narrowed down to $110^{\circ} - 125^{\circ}$ E and 15° S - 8° S. ROMS data output in netCDF file. Read data processing using Grads software to plot each variable oceanography which used for overlay to each depth position with ALB caught.

2.2.2. Fish Data

The fishery data is consisted of daily ALB catchment based number of branch line longline fishing gears, fishing effort and fishing position in latitude and longitude for the period from 2005 to 2009, obtained from scientific observer program RITF (Research Institute for Tuna Fisheries). These data were extracted to obtain geographic information where the ALB were caught, time when the operational catchment and number individual ALB which caught on each branch line of longline fishing gears as information on how the depth of ALB caught.

2.2.3. Hook Depths Estimation

In the measurement of depth of the hook, there are many factors which affect, among other things: the amount of basket that is used, total length of the main line, length of branch line, length of floatline, the speed of the mainline thrower to speed of vessel, total time for settings and current factor as a correction factor. The depth of the hook can be estimated using the formula Yoshihara (1951) as follows:

$$D = fl + bl + \frac{1}{2}BK \left\{ \sqrt{(1 + \text{Cot}^2 \sigma)} - \sqrt{(1 - 2j/n)^2 + \text{Cot}^2 \sigma} \right\} \quad (1)$$

Where :

- D : The depth of jth hook (for each basket, the two hooks closest to the floats are both numbered 1, assuming that branchlines were hung symmetrically),
- fl and bl : The length of floatline and length of branchline, respectively.
- BK : The length of mainline in unit basket,
- n : The number of branchline in unit basket.
- j : Number of hooks position
- σ : The angle between horizontal line and tangential line of the mainline at connecting points of mainline and floatline, and

$$K = \frac{Vk \times Ts}{BK \times \sum b} \quad (2)$$

Where

- K : The sagging ratio
- Vk : Speed of the mainline thrower to speed of vessel (m/hours),
- Ts : Setting time,

BK : The length of mainline in unit basket (m)

$\sum b$: Number of basket in used.

The relationship between the value of K , σ and $\sigma \text{ Cotg}2 \sigma$ can be seen in Table 1.

Table 1. The relationship between the value of K , σ and $\sigma \text{ Cotg}2 \sigma$

K	σ (0)	$\text{Cotg}2 \sigma$
0,902	40	1,2767
0,847	50	0,7038
0,769	56	0,4727
0,725	60030'	0,3300
0,703	65	0,2077
0,661	68	0,1630
0,540	72	0,1331
0,000	90	0,0000

2.2.4. Regression Through Origin

Textbooks rarely discuss RTO other than to caution against dropping the constant term from a regression, on the grounds that imposing any such restriction can only diminish the model's fit to the data. There are, however, circumstances in which RTO is appropriate or even necessary (Eisenhauer, 2003). First, RTO may be unavoidable if transformations of the OLS (ordinary least-squares) model are needed to correct violations of the Gauss–Markov assumptions. Consider, for example, the simple linear regression of Y on x . The basic linear regression model is given by:

$$Y_i = \beta_0 + \beta_1 x_i + e_i \tag{3}$$

Where β_0 is the intercept, β_1 is the slope and e_i denotes the i^{th} residual. Lagging observations and taking first differences (i.e. subtracting each observation from its successor) to correct for serial correlation in the errors requires transforming equation (26) into an RTO equation of the form:

$$Y_i - Y_{i-1} = \beta_1(x_i - x_{i-1}) + (e_i - e_{i-1}) \tag{4}$$

Alternatively, applying weighted least squares to correct for heteroscedasticity will result in a model with no intercept if the weighting factor (z) is not an independent variable. In that case, β_0 becomes a coefficient and equation (27) is replaced by a multiple linear regression without a constant:

$$Y_i / Z_i = \beta_0(1 / Z_i) + \beta_1(X_i / Z_i) + (e_i / Z_i) \tag{5}$$

Even without such transformations, however, there are often strong a priori reasons for believing that $Y = 0$ when $X = 0$, and therefore omitting the constant. Final the simple linear RTO model equation is:

$$Y_i = \beta_1 x_i + e_i \tag{6}$$

Where β_1 is the slope and e_i denotes the i^{th} residual. Two hypothesis apply to determine linearity of slope were stated below:

$$H_0 : \beta = 0 \tag{6}$$

$$H_1 : \beta \neq 0 \tag{6}$$

Single parameter t-test was used to test those hypothesis. The t-value resulted from linear regression output calculated by dividing estimated value of predictor variable with it's standard error. The t-calculated value compared with with critical t-value from t-distribution table with 95% confidence interval. If the t- calculated value less then t-distribution table, the H_0 rejected. It's mean there is linear relationship between response variable and predictor variable (Vezani, 2005; Zuur et al., 2007).

In addition to linear regression, p-value was used to measure the linear relationship between response variable and predictor variable. p-value is the probability that test the statistic test on the observed values (Verzani, 2005). If a p-value is small the test is called statistically significant, as it indicates that the null hypothesis is unlikely to produce more extreme values than the observed one. The p-value range is described on the Table 2.

Table 2. Level of significant for range of p-values (Verzani, 2005)

p-value range	Common description
<0.001	extremely significant
0.001-0.01	highly significant
0.01 - 0.05	statically significant
0.05 - 0.1	could be significant
0.1 - 1	not significant

The coefficient determination (R^2) was used to validate the regression linear model. The higher value of R^2 , the more the model can be explained (Zuur et al., 2007). The R_0^2 calculated from RTO the given equation below:

$$R_0^2 = \frac{\sum y_i^2}{\sum y_i^2} \quad (7)$$

Where:

R^2 : Coefficient determination

\hat{y} : fitted y value

\bar{y} : mean of y

When R^2 close to 1, most of the variation is explained by the regression line, and when R^2 close to 0, regression line does not much explain the variation (Verzani, 2005). All proses statistical using R software. In order to choose the final linear model equation, the p-value, Akaike information Criteria (AIC), residual deviance explained of the lm outputs were considered. In R software the criterion used is:

$$AIC = -2 * \text{Log}L + k * \text{edf} \quad (8)$$

Where L is the likelihood and edf the equivalent degrees of freedom (i.e., the number of free parameters for usual parametric models) of fit. For linear models with unknown scale (i.e., for lm and aov), $-2 \log L$ is computed from the deviance and uses a different additive constant to logLik and hence AIC.

3. Results

3.1 Vertical Distribution of ALB

ALB fishing ground was mapped base on depth catchment data (Figure 2). Color dots on the image shows position and depth of ALB caught. The data were divided by month during 2005-2009, unfortely there is no data in January and February, this is due to extreme weather conditions causing no ship activity in Benoa port. ALB fishing grounds is mostly done outside the EEZ, because of ALB is high migratory and oceanic spesies.

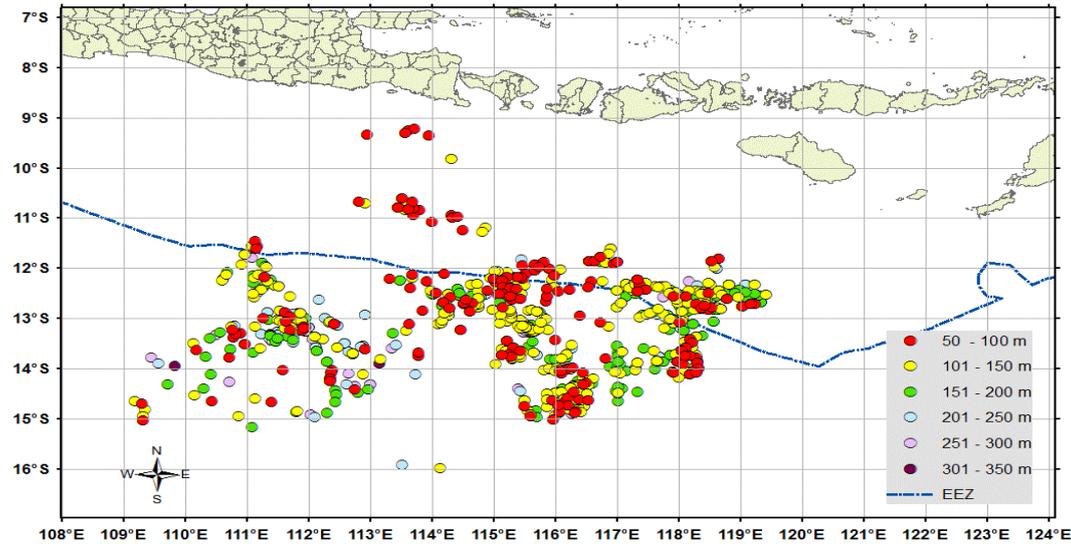


Figure 2. Distributions of ALB caught (without January and February)

Vertical distributions divided into 6 layers of depth of water it was decided based on the depth of fishing gear longline which used for ALB can reach a depth of 450m. Based on calculation of Yoshihara formula, vertical distributions of ALB from the depths of 50-400 meters where ALB mostly were caught at a depth of 101-150 meters and decreased number of catches directly proportional with increasing depths (Figure 3).

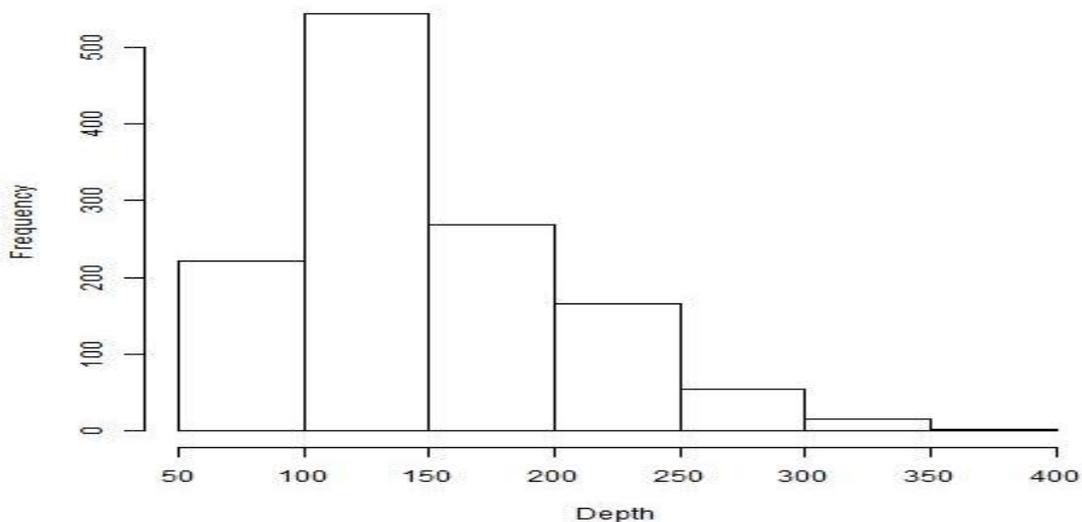


Figure 3. Frequency of ALB catches by the depth of hooks

3.2 Oceanographic Variabels on ALB Fishing Gorund

In addition to estimate distribution ALB based on oceanographic preferences during period 2005-2009, frequency value of each oceanographic variable was performed using histogram. The pattern of distributions of ALB caught in the fishing ground indicated that temperature range between 13.28 – 26.74°C during this period (2005-2009). Wherein distribution patterns formed by swimming layer of ALB. There are 6 groups with a temperature range from distribution pattern of ALB, i.e. : 13°C; 15°C; 17°C; 19-20°C; 22-23 °C and 26°C respectively. The percentage of temperature wherein ALB caught i.e. : 23°C (24.45 %), 22°C (17.43 %), 19°C (14.20 %) 26°C (13.56 %) and 17°C (10.41 %) respectively, while the percentage of other temperature below 10% (Figure 4).

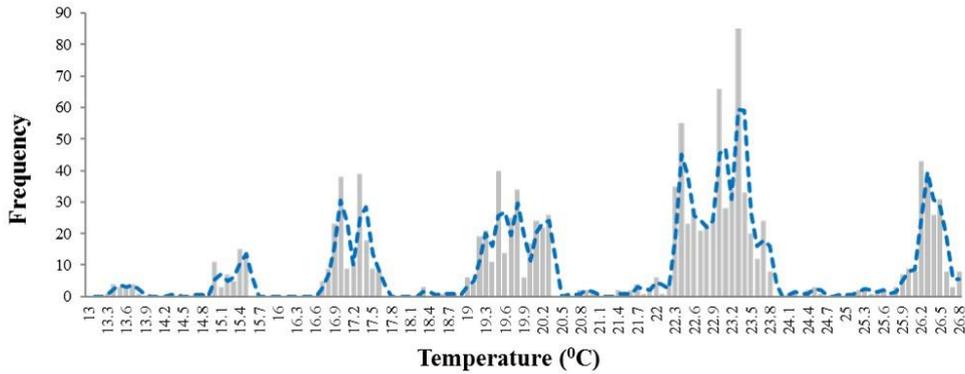


Figure 4. Temperature distributions of ALB caught during 2005-2009 period

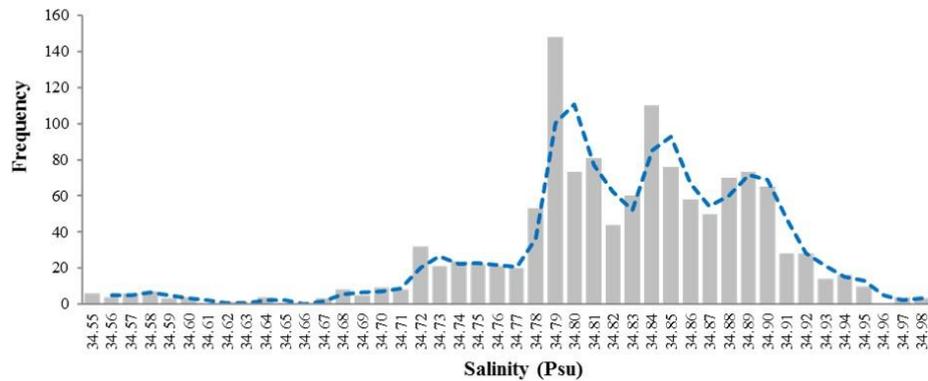


Figure 5. Salinity distributions of ALB caught during 2005-2009 period

Figure 5, showed the pattern of ALB distributions based on Salinity that is not too varied. The range of salinity values at point 34.55 to 34.97 Psu. The range point of salinity showed a very thin difference with a scale 0:01 Psu, wherein ALB mostly caught at salinity range on 34.79 - 34.98 Psu. The percentage of salinity wherein ALB caught is 34.79 (11.64 %), 34.84 (8.65 %) and 34.81 (6.37 %) respectively, while the percentage of other salinity below 6 %.

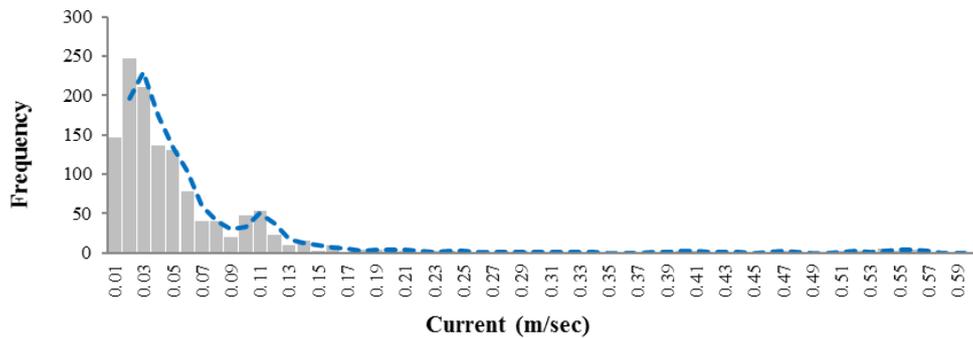


Figure 6. Current velocity distributions of ALB caught during 2005-2009 period

Figure 6, showed current velocity was 0.01 to 0.56 m/sec. ALB mostly caught with current velocity range average 0.01 to 0.06. current velocity conditions very low caused the position of fishing grounds quite far from The Indonesian through flow and mostly ALB swimming layer depths occurred on 100 – 200 meter. The percentage of current velocity wherein ALB caught is 0.02 m/s (19.46 %), 0.03 m/s (16.63 %), 0.01 m/s (11.51 %), 0.04 m/s (10.72 %) and 0.05 m/s (10.24 %) respectively, while the percentage of other current velocity below 6 %.

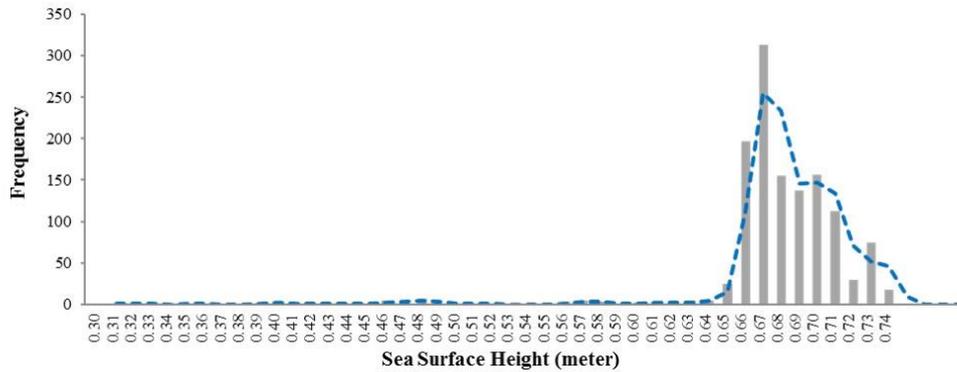


Figure 7. SSH distributions of ALB caught during 2005-2009 period

Figure 7, showed SSH conditions ALB caught were 0.31 – 0.74 meter. The range point of SSH showed a very thin difference with a scale 0:01 meter. ALB mostly caught with SSH range average 0.65 to 0.74 meter. The percentage of SSH wherein ALB caught is 0.67 (24.68 %), 0.66 (15.54 %), 0.70 (12.30 %), 0.68 (12.22 %) and 0.69 (10.80 %) respectively, while the percentage of other SSH below 10 %.

3.3 Linear Regression Result

Prior to examining the relationship between ALB catches and environmental variables, we examined the relationship between number of ALB caught and each oceanographic variables. The effect of environmental conditions, deduced from linear regression outputs, indicated that environmental variables strongly influenced the numbers of ALB caught.

Table 3. showed list regression model variable, p-value, t value, and R^2 . The predictor variables were highly significant ($P < 0.01$). Regression model showed that the critical value t-value of oceanographic data is 1.66 (significance level is 0.05, two sided, $df = 67$). Based on the t-value, all predictor variable has linear relationship with response variable. The t-value all variables was higher than critical t-value. It is mean, the linear regression of the variable has $\beta \neq 0$.

Table 3. Results of regression model for each parameter

Response Variable	Predictor Variables	p-value	t value	R02
Number of ALB based on depth	Temperature (0C)	1.69E-14	9.725	0.5756
	SSH (m)	3.88E-13	8.967	0.5418
	Salinity (PSU)	4.32E-13	8.94	0.5336
	Current (m/s)	2.46E-09	6.867	0.4008

3.4 Model Selection

Before determining the model to be used in this study, several types of test procedures have been performed to test goodness of fit model. All test procedure performed in R software. Pairplots used to determine whether there is collinearity (interplay) between oceanographic variables were examined. Collinearity is one of the requirements for regression analysis. Finally, only variable depth, temperature and salinity cannot be used as predictor variable at the same time, so there are only 3 possibilities to establish a regression equation i.e:

$$lm(\text{formula} = \text{ALB} \sim \text{Depth} + \text{Curr} + \text{SSH} - 1) \quad (9)$$

$$lm(\text{formula} = \text{ALB} \sim \text{Temp} + \text{Curr} + \text{SSH} - 1) \quad (10)$$

$$lm(\text{formula} = \text{ALB} \sim \text{Sal} + \text{Curr} + \text{SSH} - 1) \tag{11}$$

The Akaike information criterion (AIC) value were used to determine the best model, the best model selected based on Smallest value of AIC. From three model linear regression, model with formula: $lm(\text{formula} = \text{ALB} \sim \text{Depth} + \text{Curr} + \text{SSH} - 1)$ is the level of the highest goodness of fit with higher R_0^2 and smaller AIC than other models. The final Regression equation for goodness of fit was:

$$y = -0.06024 \times \text{Depth} - 118.62402 \times \text{Curr} + 40.52377 \times \text{SSH} \tag{12}$$

Where :

- y : Number of ALB (n)
- Depth : Swimming layer ALB caught (m)
- Curr : Current velocity (m/s)
- SSH : Sea Surface Height (m)

3.5 Model Validation

Model validations were showed on Figure 8. Residual vs Fitted which has better than other predictor variables which equally spread residuals around a horizontal line without distinct patterns, that is indicated regression line has explain higher linear relationships. Normal Q-Q used checks how closely the data follow a normal distribution. all data has residuals follow a straight line well so that it can be said the assumption of normality is fulfilled. Scale-Location, this plot shows if residuals are spread equally along the ranges of predictors. This graph used to check the assumption of equal variance (homoscedasticity). from all model the residuals appear randomly spread. It's good pattern when horizontal line with equally (randomly) spread points. Residual vs Leverage, this plot helps us to find influential cases if any. Not all outliers are influential in linear regression analysis (whatever outliers mean). Even though data have extreme values, they might not be influential to determine a regression line. That means, the results wouldn't be much different if we either include or exclude them from analysis. They follow the trend in the majority of cases and they don't really matter; they are not influential. On the other hand, some cases could be very influential even if they look to be within a reasonable range of the values. They could be extreme cases against a regression line and can alter the results if we exclude them from analysis. Another way to put it is that they don't get along with the trend in the majority of the cases. In this graph there is no influential case, or cases in each model, caused can barely see Cook's distance lines (a red dashed line) because all cases are well inside of the Cook's distance lines.

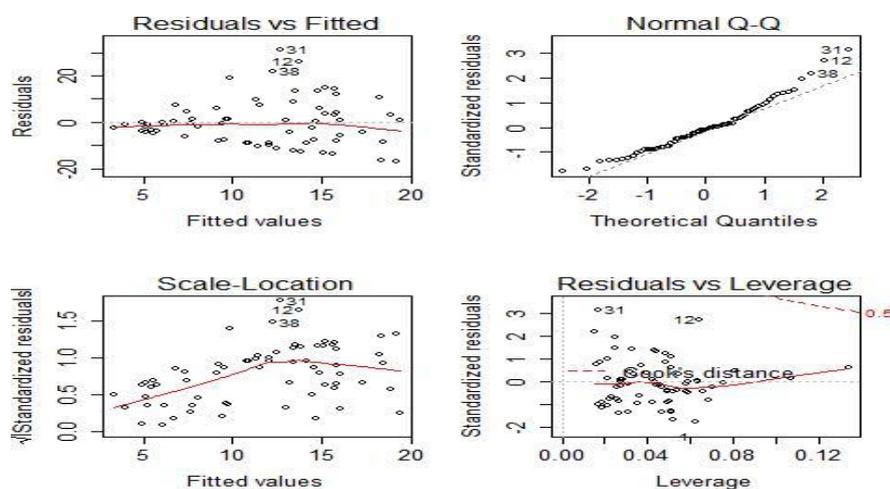


Figure 8. Model validations linear regression

4. Discussion

Fishing ground of ALB mostly done outside the EEZ, this was due to ALB is high migratory and oceanic species and the same distribution each month (Figure 2). Wudianto et al. (2003), stated fishing ground longliners from Cilacap and Benoa port are South central of Java island between 108° - 118° E and 8° - 22° S and fishing operation of longliners (>70%) was mostly outside the EEZ of Indonesia. There are two zones of longliners fishing ground base on Benoa port position, ie the zone in the southeastern (south and east) and the zone of southwestern (south and west), intensity of longliners more focused operating in the area southwestern of the Benoa port wherein this area is a potential fishing ground for frozen tuna fleet (Novianto et al. 2010).

Hendiarti et al. (2005) concluded that the “pelagic fish season” develops during the southeast monsoon where maximum and minimum peaks during Jun.-Sep. and November - January respectively. Furthermore, Susanto & Mara (2005) stated during the southeast monsoon (April to October), Southeasterly wind from Australia generates upwelling, bringing cooler waters and increased nutrients to the surface along the southern coasts of Java and Sumatra. Conditions are reversed during the northwest monsoon (October to April).

The results showed southeast monsoon (June to August) catch data showed that the abundance of ALB was higher rather than in another monsoons. This phenomenon allegedly related with occurrence of upwelling in this region. This result was similar with Lan et al. (2011) indicated that the major fishing period in indian ocean is winter (June to September), and the high CPUEs occur in this period, while low CPUEs occur in summer (March to May), meanwhile Conand et al. (2007) revealed that in the Reunion Island part of West Indian ocean, peak seasons ALB fishing occurred in November-January (transitional warming season). On different location, Briand et al. (2011), revealed that a large part of ALB CPUE variability can be explained by seasonal, interannual and spatial variation of the habitat.

Swimming layer is one of the important factor to get maximum catches, especially on tuna longline effort. ALB, being negatively buoyant in seawater, must swim continuously to maintain their position in the water column (Dotson, 1976). Based on spatial distribution analysis ALB mostly were found in the depth around 100 - 150 m (Figure 5.2). This result was almost similar with Domokos et al. (2007), who was mentioned that ALB spend most of their time between 150 and 250 m during the daytime and between 0 and 200 m during the night-time, at depths coinciding with those of small local maxima in micronekton biomass whose backscattering properties are consistent with those of the preferred prey of ALB. Arrizabalaga et al. (2015) stated ALB preference and tolerance ranges extend to about 200 m. Meanwhile Nugraha & Triharyuni (2009) mentioned that ALB in Indian Ocean were caught in depth of 150 - 199.9 m, were also comparable to those in Sukresno et al. (2014) in the same region concluded ALB were found in the depth around 150 m. Williams et al. (2014) revealed evidence that ALB tagged in the two tropical locations rarely occupied water shallower than 100 m during the day, remaining below the MLD, with fish spending most (New Caledonia: 80%, Tonga: 88%) of the time between 150 and 250 m and 200–350 m, respectively and spent most (New Caledonia: 56%, Tonga: 63%) of the night-time between 50 and 150 m and Maximum depths recorded by ALB tagged in New Caledonian, Tongan and New Zealand waters were 372 m, 409 m and 250 m, respectively.

Analysis Sub-surface sea water temperature showed fluctuates seasonally. Monthly average of sub-surface temperature was found tend to vary from January to December. Qu et al. (2005), who had stated that sub-surface temperature in Indian Ocean was influenced by monsoon. In addition to sub-surface temperature, the range of sub-surface temperature during northwest monsoon in the depth around 51-100 m was about 29.20 °C, while in southeast monsoon was about 27.94 °C. Temporal distribution of sub-surface temperature in Indian Ocean part South of Java was warmer in northwest monsoon than in southeast monsoon (Sukresno et al., 2014; Hartoko, 2010). Variability oceanographic condition maps were shown on Appendix 2.

This study shows that most caught ALB in temperature conditions ranging between 22-23 °C with the proportion of total 532 (41.82%) of the total 1272 ALB caught. Different result from Sukresno et al. (2014) in the same region concluded ALB were found with

optimum temperature between 17° – 21° C, where ALB avoids waters below 14° C (Arrizabalaga *et al.*, 2015). Meanwhile Zainuddin et al, (2011) in the western North Pacific Ocean indicated that the highest concentrations of ALB were found in water of 18.5 - 21.5° C and tended to be centered at 20° C. Williams et al. (2014) found evidence that ALB tagged in New Zealand waters primarily spent most (85–97%) of their time in water 17 – 20 °C both during the day and during the night, and in the two tropical locations, ALB spent most of their time (New Caledonia: 80%, Tonga: 98%) at night between 22 and 26 °C, with the fish tagged in Tongan waters occupying mostly cooler waters (16–21 °C) during the day than those ALB tagged off New Caledonia (19–23 °C).

The salinity of the former is less than that of the latter, with range of the ALB closely correlated to catch rate is from 34.68 to 34.95 PSU, and the range more closely correlated to catch rate is from 34.72 to 34.92 PSU with the proportion of total 1,157 (90.95 %) of the total 1272 ALB caught. This result was almost similar with Roberts (1980), stated that most ALB were caught either in subtropical water or in the Subtropical Convergence Zone with range of salinities 34.6 - 35.4 ‰ in New Zealand water. Different result from Arrizabalaga *et al.* (2015) which states that ALB and skipjack showed clearly defined preferred salinity ranges, at around 36-37 PSU. Chen et al. (2005) concluded that Higher CPUE of ALB occurred in Salinity 35.32 PSU, and concentration of immature ALB matched the higher salinity zones, yet they could not conclude from their observations whether the current system or the salinity itself caused the coincidence.

ALB is a temperate to subtropical species with highest CPUEs between 20°-40° of each hemisphere (Arrizabalaga *et al.*, 2015). The distribution pattern of ALB in the Indian Ocean for imature and mature ALB tuna based north-south seasonal migration, within 10° and 30° S, the separation of mature, spawning, and immature ALB life history stages roughly coincided with the boundaries of the three oceanic current systems in the Indian Ocean (Chen *et al.*, 2005). The favourable ALB fishing grounds are located along the oceanic frontal region, higher hook-rate of ALB were found around or adjacent to the oceanic fronts, i.e. around the water boundary of the equatorial countercurrent and south equatorial current; near the water boundary (convergence) between the south equatorial current and the Indian Ocean central water; near the subtropical convergence in the southern hemisphere. (Uda & Nakamura, 1973). Xu et al. (2013), revealed in the open ocean, high CPUE ALB is related with a small to moderate GCM (absolute values are <0.15m·s⁻¹). Current velocity in the area ALB caught very low, mostly ALB caught were current velocity on 0.02 m/s with the proportion of total 248 (19.46 %) of the total 1272 ALB caught. Allegedly at the ALB fishing ground area there are two types of current i.e., South Equatorial current (SEC) and Rossby Waves (RW) that affect the abundance of ALB. The waves affect the thermocline depth, and hence the ocean's temperature profile, as well as the sea surface height and can either raise or lower the thermocline, leading to a decrease or in crease respectively in the sea surface height (Gnanaseelan et al., 2004).

Tropical tunas, as well as ALB, prefer slightly negative or positive values of SSH (Arrizabalaga *et al.*, 2015). According to Xu et al. (2013), in the open ocean, SSH anomaly is an important indicator of ALB habitat. Negative SSH anomaly is associated with cyclonic eddies and divergent upper ocean waters. Under these conditions, cool and nutrient rich waters may be upwelled from deeper ocean as compensation to the divergent surface layer. These upwelling eddies are areas of increased productivity in the open ocean and may attract various prey sources for ALB. These studies show that SSH was more important than temperature, salinity or current in the study area. Based on goodness fit multiple regression model SSH having the highest t-value in all models with higher significance level than other predictor variables. ALB mostly caught with SSH range average 0.65 to 0.74 meter. The percentage of SSH wherein albakor tuna caught were occurred on the boundary 0.67 m with the proportion of total 314 (24.68 %) of the total 1272 ALB caught.

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