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Standardized CPUE of Japanese commercial dip-net fishery targeting spawners of chub mackerel in the Northwest Pacific up to 2023

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Summary

- We conducted CPUE standardization of Japanese commercial dip-net fishery for Pacific chub mackerel using a generalized linear mixed-effect model
- The analysis showed that the dip-net fishery CPUE was affected by month, area, sea surface temperature, and ship as well as year.
- The abundance index standardizing these influential variables except for year showed a great decline in 2022-2023 after a high-level decade from 2011 to 2021.
- We propose this standardized index to be used as an index of spawning stock biomass (SSB) in the Technical Working Group for the Chub Mackerel Stock Assessment (TWG CMSA) in NPFC.

Background

- The dip-net fishery operating around the Izu islands is a small-scale artisanal fishery targeting spawning chub mackerel during the spawning season
- The total catch amount of chub mackerel in this fishery contributes less than 1% of the overall catch by Japan (Table 1)
- It is the only fishery that targets spawning chub mackerel and operates in the main spawning ground around the Izu Islands during the spawning season
- Most mature fish are considered to migrate to this area for spawning (Watanabe and Yatsu 2006)
- The CPUE of the dip-net fishery is considered to represent the relative abundance of spawning stock biomass (SSB) for the Pacific chub mackerel
- The CPUE has long been used as a reliable abundance index of SSB in the Japanese domestic stock assessment.

Changes since the last document

	Previous WP	Current WP		
Model	GLM	GLMM		
Probability distribution	Hurdle (delta) Gamma	Zero-inflated Tweedie		

- A previous working paper used GLM (Nishijima et al. 2022), while this working paper used GLMM
- GLMM estimated the year effect in the binomial model (encounter probability) as random effects because some years had only positive catch samples and using fixed effects caused extremely high values of parameter estimates and uncertainties
- We changed the probability distribution used from hurdle (delta) Gamma to Zero-inflated Tweedie because of much reduced AICc

	Hurdle-	Hurdle-	'Normal'	Zero-inflated
	lognormal	gamma	Tweedie	Tweedie
ΔΑΙϹϲ	294.53	290.53	426.28	0.00

Catch and effort information

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	Number of	% Coverage of	% Coverage of	Total Catch	Total Effort for	Percentage of	
Year	observations ¹	CPUE FLEET	CPUE FLEET	CPUE FLEET	CPUE FLEET	overall catch by	
	observations	$(\text{catch})^2$	(effort) ²	(mt)	(fishing days) ³	member $(\%)^4$	
2003	132	21.56	20.99	60.81	132	0.13	
2004	168	25.96	27.57	41.68	166	0.06	
2005	117	26.64	25.16	34.97	117	0.02	
2006	117	46.38	49.16	143.86	117	0.06	
2007	198	14.91	43.14	350.95	198	0.14	
2008	104	28.46	13.22	124.77	103	0.07	
2009	112	31.22	31.73	137.79	112	0.08	
2010	118	36.34	39.33	124.15	118	0.10	
2011	105	22.57	43.15	177.01	104	0.14	
2012	76	10.05	44.44	49.99	76	0.05	
2013	98	21.01	26.98	495.18	58	0.39	
2014	117	25.06	30.93	723.53	73	0.33	
2015	84	36.59	32.70	851.09	52	0.30	
2016	129	31.37	26.81	1492.43	85	0.45	
2017	124	55.52	38.30	537.62	72	0.16	
2018	113	33.38	26.84	1194.23	73	0.36	
2019	120	45.37	32.43	1436.21	84	0.48	
2020	178	47.64	37.06	1980.79	106	0.74	
2021	179	44.69	33.44	1467.18	104	0.53	
2022	72	36.31	22.61	549.65	52	0.29	
2023	88	41.25	24.66	253.88	73	0.27	

Table 1

- The data of dip-net fishery from 2003 to 2023 was obtained from the logbooks from eight sampling ships in Kanagawa and Shizuoka Prefectures
- The coverage of catch from the sampling ships against the total catch of the dip-net fishery is 10 to 56%
- The data was recorded by operation by ship, along with the information on locations (longitude/latitude or area name), *in-situ* sea surface temperatures (SST), the number of fishermen (nets), and fishing time

Filtering rule

Eilton Applied	Number of Records	Number	Number of Records with
Filter Applied	Remaining	Removed	Chub Mackerel Catch >0
Initial Data set	2,549	-	1,991
Remove data with no spatial	2 520	10	1 097
information (area or long/lat)	2,539	10	1,987
Remove data with no effort (time	2 424	105	1 000
and person)	2,434	105	1,909
Remove data with $SST = NA$ or 0	2 416	18	1 800
(not recorded)	2,416	18	1,899
Select data between January and	2 2 4 2	174	1 820
June	2,242	1/4	1,829



- The number of samples in the original data was 2,549
- We removed data with no spatial information, data with no effort information (fishing time and the number of fishermen), and no SST information from the analysis
- We exclusively focused on the data from January to June, the main spawning season of chub mackerel, and removed the data obtained during the other months.
- The sample size of the final dataset was 2,242 and that having positive catch was 1,829 (81.6%)

Area division, and relationship between area and effort



- The dip-net fisheries are conducting in the area approximately from 138°-140.5° E and 32.5°-35° N
- There are many samples that had either longitude/latitude or area name
- We assigned the area whose center was closest, to each sample that had only longitude and latitude, and then used area as a categorical variable in CPUE standardization.

Month

Catch and CPUE by area by month in each year

Fig. 2C





- The catch and CPUE were low until 2006, but have subsequently increased.
- The catch and CPUE tended to be high from February to April

Spatial maps of catch and effort

Plotted spatial maps of catch and effort only for data having information on longitude and latitude



Spatial map of CPUE

Plotted spatial map of CPUE (kg/net-hour) only for data having information on longitude and latitude

Fig. 3C



Explanatory variables used

Table 3

Variable	Abbreviation	Number of category	Detail	Note
Year	year	21	2003-2023	Treated as fixed effect for Tweedie and as random effect (AR1) for the binomial distribution
Month	month	6	January-June	Categorical variable with fixed effect
Area	area	7	A-G	Categorical variable with fixed effect
Sea surface temperature	SST	-	13.2-28.2	Continuous variable scaled by mean and SD
SST squared	I(SST^2)	-	Squared SST	Squared values of the scaled SST
Prefecture	pref	2	Belonging of ship (Kanagawa or Shizuoka)	Categorical variable with fixed effect
Ship	ship	8	Sampling ship	Categorical variable with fixed effect

- All variables except for SST and its squared term were categorical variables
- The effect of year was treated as fixed effect for Tweedie distribution and as random effect for binomial distribution assuming AR(1) process
- This is because several years have only positive catch samples

Associations of year, month, ship, and prefecture

The variables of ship and prefecture

operations strongly depended on

There are many missing categories

prefectures

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have a nested structure and year with



Associations of year, month, and area



Allocations of efforts to each area depended on the years

Correlation among the categorical variables

Fig. 4C



Cra	amer's 1.00	١
	0.75	
	0.50	
	0.25	
	0.00	

Cramer's V (like correlation coefficient for categorical variables) was

- High between prefecture and ship and between prefecture and year (>0.9)
- Moderately high between year and area, between year and ship, and between area and prefecture (0.4~0.5)

Relationships between SST and categorical variables





Fig. 4D

- SST was strongly correlated with month
- There was no apparent correlation of SST to the other categorical variable

Relationships between CPUE and categorical variables





Fig. 4E

CPUE was seemingly correlated by all the categorical variables

Full model description and model selection method

- The dependent variable CPUE (kg/net-hour) was a continuous value more than or equal to zero
- Used a generalized linear mixed-effects model (GLMM) with a zero-inflated Tweedie distribution via the R package 'glmmTMB' (Brooks et al. 2017).
- The zero-inflated Tweedie distribution in this study is a mixture of binomial distribution (with logit link) and Tweedie distribution (with log link).
- The full model involved all the five categorical variables (year, month, area, prefecture, and ship)
- Considered the squared term of SST in the full model because CPUE seemed to be the highest at an intermediate level of SST
- Not consider interactions between any combination of the independent variables because including interactions would cause many missing categories
- Estimated all parameters as fixed effect except for the year effect in the binomial model
- Conducted the brute-force model selection approach except that the year effect was always selected and models with both prefecture and ship were not considered because of their nested structure and the strongest correlation (Fig. 4A, C),
- Based on AICc useing the R package 'MuMIn' (Bartoń 2022)

Model selection results

Table 4

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Rank	(T)area	(T)month	(T)pref	(T)ship	(T)SST	$(T)I(SST^2)$	(T)year	(B)area	(B)month	(B)pref	(B)ship	(B)SST	(B)I(SST^2)	df	logLik	AICc	ΔAICc
1	+	+		+	0.116	-0.117	+	+	+			0.731	0.191	59	-10137.14	20395.53	0
2	+	+		+	0.115	-0.117	+	+	+	+		0.720	0.190	60	-10136.78	20396.91	1.38
3	+	+		+	0.116	-0.122	+	+	+			0.774		58	-10140	20399.13	3.6
4	+	+		+	0.115	-0.122	+	+	+	+		0.756		59	-10139.68	20400.6	5.07
5	+	+		+	0.115	-0.117	+	+	+		+	0.690	0.190	66	-10135.04	20406.15	10.62
6	+	+		+	0.115	-0.121	+	+	+		+	0.729		65	-10137.98	20409.9	14.37
7	+	+	+		0.139	-0.113	+	+	+			0.741	0.197	53	-10152.3	20413.23	17.7
8	+	+	+		0.139	-0.113	+	+	+	+		0.728	0.195	54	-10151.98	20414.68	19.15
9	+	+		+	0.156		+	+	+			0.727	0.211	58	-10148.18	20415.5	19.97
10	+	+		+	0.156		+	+	+	+		0.719	0.208	59	-10147.78	20416.8	21.27
11	+	+	+		0.139	-0.118	+	+	+			0.775		52	-10155.27	20417.07	21.54
12	+	+	+		0.139	-0.118	+	+	+	+		0.762		53	-10154.99	20418.6	23.07
13	+	+		+	0.159		+	+	+			0.770		57	-10151.92	20420.87	25.34
14	+	+		+	0.158		+	+	+	+		0.751		58	-10151.51	20422.15	26.62
15	+	+		+	0.101	-0.125	+	+	+					57	-10153.36	20423.74	28.21
16	+	+	+		0.139	-0.112	+	+	+		+	0.703	0.196	60	-10150.3	20423.95	28.42
17	+	+		+	0.101	-0.125	+	+	+	+				58	-10152.82	20424.78	29.25
18	+	+			0.129	-0.112	+	+	+			0.757	0.196	52	-10159.57	20425.67	30.14
19	+	+		+	0.155		+	+	+		+	0.688	0.208	65	-10146.05	20426.04	30.51
20	+	+			0.129	-0.112	+	+	+	+		0.744	0.195	53	-10159.36	20427.34	31.81

- The effects of area, month, and SST were always selected in both Tweedie and binomial parts in the top 20 models
 - Squared SST was also selected for both distributions in the top model with minimum AICc.
 - Ship was selected only in binomial distribution in the top model.
 - Selected the model with minimum AICc as the base model.

Analysis of deviance table

Table 5

Variable	Chisq	Df	Pr(>Chisq)		Significance	% deviance
v allaule	Chilsq	DI		T (>CIIIsq)	code	explained
<u>Tweedie</u>						
area	315.28		7	3.30E-64	***	8.64%
month	159.99		5	9.96E-33	***	
ship	46.45		7	7.15E-08	***	
SST	8.43		1	3.68E-03	**	
I(SST^2)	22.53		1	2.07E-06	***	
year	311.50		20	3.60E-54	***	
<u>Binomial</u>						
area	72.13		6	1.49E-13	***	
month	42.46		5	4.75E-08	***	
SST	21.00		1	4.58E-06	***	
I(SST^2)	5.93		1	1.49E-02	*	

- The effects of area, month, and SST had significant influences on CPUE in both Tweedie and binomial parts in the base model, according to the likelihood ratio test using the chi-square statistic
- The percent deviance explained of the base model was 8.64%

Model diagnostics for scaled residuals

- Generated scaled residuals using the R package 'DHARMa' (Hartig 2022) for model diagnostics
- This package enables to simulate the scaled residuals which should theoretically follow the uniform distribution from zero to one





Relationship between explanatory variables and predicted CPUE

Fig. 6: Partial dependence plot



- CPUE was expected be the highest at 19.4°C
- CPUE was higher in February to April than in the other months.

Yearly trends of nominal and standardized CPUE

- To derive the standardized CPUE values, we calculated predicted CPUE values per each category (for the continuous variables, we divided their range at small regular intervals) of selected variables (e.g., Area = A, B, C..., Year = 2003, 2003, 2004..., SST = 10.0, 11.0, 12.0...) using the *expand.grid* function in R
- Then calculated the arithmetic mean of each year.
- Not implement an area-weighting approach because the size of each area was unknown.



- Standardized CPUE has been relatively low until 2005, increased since then, and remained relatively stable at a high level from 2011 to 2021
- However, it declined significantly thereafter and was at its lowest in 2023 since 2006.
- This yearly trend of the standardized CPUE was not largely different from that of nominal CPUE except that the scaled standardized CPUE was much lower in 2020 than the scaled nominal value

Why nominal and standardized values were much different in 2020?

- The samples in 2020 had a large proportion of February to April and ship IDs of 3 and 4, when CPUE tends to be higher,
- There were no operations in areas B and D, when CPUE tends to be lower, which elevated the nominal CPUE





Values and uncertainties of the nominal and standardized CPUE

Table 7	Year	Nominal	Nominal Standardized		Lower	Upper
Table 7		(kg/net-hour)	(kg/net-hour)	CV	95% CI	95% CI
	2003	5.49	3.48	0.22	2.25	5.51
	2004	4.46	5.45	0.2	3.84	8.32
	2005	3.29	2.12	0.26	1.32	3.61
	2006	25.46	12.99	0.22	8.73	19.65
	2007	86.56	36.84	0.12	29.71	47.69
	2008	45.53	15.94	0.15	12.28	22.05
	2009	56.51	24.48	0.15	18.72	33.57
	2010	54.51	22.43	0.15	17.19	30.91
	2011	116.21	42.21	0.15	32.44	57.76
	2012	120.54	45.06	0.16	33.5	62.61
	2013	131.91	52.92	0.28	31.77	92.3
	2014	110.94	40.53	0.16	30.57	55.76
	2015	120.32	37.3	0.17	27.49	54.01
	2016	172.48	62.71	0.15	48.09	85.24
	2017	81.48	31.59	0.17	23.29	45.5
	2018	142.86	53.44	0.16	40.61	74.43
	2019	142.44	44.88	0.13	35.82	58.58
	2020	167.34	44.58	0.14	34.84	60.1
	2021	115.21	40.56	0.15	31.22	57.09
	2022	63.17	18.86	0.18	13.55	27.32
	2023	23.91	7.17	0.19	5.09	10.46

The coefficients of variation (CV) of the estimates were 0.14-0.28

Discussion and recommendation

- The dip-net fishery CPUE was influenced by the factors of month, area, in-situ SST, and ship
- These factors were considered to have an impact independent of the stock abundance in each year, and hence, standardized to eliminate sampling biases
- The standardized index values showed a relatively stable trend at high levels from 2011 to 2021, followed by a sharp decline in 2022 and 2023.
- In terms of model diagnostics, issues such as scaled residuals deviating from theoretical values were observed, and the % deviance explained was low.
- This might be attributed to the considerable variability in the original data, imperfect spatial information, and the possibility of overlooking other important variables such as interactions among explanatory variables
- There might be room for model improvement in the future
- It is believed that the majority of spawning chub mackerel migrates around the Izu Islands and, therefore, the CPUE of the dip-net fishery targeting the spawners represents valuable information based on the direct observations of spawning fish of chub mackerel.
- We propose to use the standardized CPUE values calculated in this study as an abundance index of SSB in CMSA.