ORIGINAL RESEARCH



Prediction of fillet yield in hatchery populations of Asian sea bass, *Lates calcarifer* (Bloch, 1790) using body weight and measurements

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Abstract The purpose of this study was to develop and validate prediction equations for fillet traits in Asian sea bass. Ninety-day old fish (average weight = 9.48 ± 0.35 g; length = 8.78 ± 0.6 cm) from four hatchery stocks (Chachoengsao, Chon Buri, Chumphon, and Samut Songkram) were raised in earthen ponds. After 300 days, weight and body measurement data of live fish (n = 400) were collected. Mean individual weight was greatest for the Chacheongsao stock (1166.32 ± 23.42 g) and was similar for the other stocks, ranging from 982.96 ± 25.07 to 997.44 ± 24.71 g. Fillet percentage varied slightly from 47.33 to 49.88%. Positive high correlations were observed for weight and body measurements with fillet weight, whereas there were significant but weak correlations for body measurements with fillet yield. Prediction equations developed from body weight using simple linear regression models yielded R^2 values of 0.97-0.98 for fillet weight for each stock. Correlations between values predicted from the body weight models and actual values were 0.98-0.99 for fillet weight. Stepwise regression was performed to develop prediction models for fillet yield from body measurements. The best fillet yield prediction models identified length and an additional 2-3 measurement as potential predictors depending on the population. Prediction biases were close to zero despite low to moderate (r values of 0.20-0.51) degrees of predictive power of the models. However, prediction models for fillet yield should be further developed to increase predictability and be applicable to new data.

Keywords Barramundi · Fillet yield · Stepwise regression · Predictive power · Value-added fish products

Introduction

Asian sea bass, also known as barramundi (*Lates calcarifer*), is the predominant fish species used for coastal aquaculture in Thailand. It is a significant fishery commodity representing 10% of all seafood production. Asian sea bass production increased by 51% from 8004 metric tons in 2001 to 16,501 tons in 2014 as reported by the Thailand Department of Fisheries (DOF 2016). Farmed sea bass is normally sold as whole fish at 500–700 g for local markets, with a retail price of 150 baht/kg (4.50 USD/kg). The consumer preference is increasingly leaning towards value-added and convenience products, such as frozen fish and fish fillets.

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Therefore, the production of larger fish (1000–2000 g) has increased during the past decade. These types of products are also exported to Taiwan, Singapore, Malaysia and Australia (Food and Agriculture Organization 2017). Value addition is one of the approaches expected to increase profitability of the sea bass farming and processing industries (Department of Fisheries 2016).

Improvement in production traits, including growth performance and yield, is considered the most important breeding goals in fish farming (Gjedrem 1997, 2000; Kause et al. 2002; Thodesen et al. 2012). Several studies have been conducted comparing growth performance and fillet percentage for various fish strains utilizing body weight and volume measurements, as well as body shape, e.g., Nile tilapia *Oreochromis niloticus* (Rutten et al. 2004; Rutten et al. 2005; Peterman and Phelps 2012), striped bass *Morone* spp. (Bosworth et al. 1998), common carp *Cyrpinus carpio* (Cibert et al. 1999) and catfish *Ictalurus* spp. (Argue et al. 2003; Bosworth et al. 2001). Fillet yield, an economically important trait for the fish processing industry, however, is difficult to measure with high accuracy without sacrificing the animal and cannot be used as a criterion during broodstock selection. A more practical approach is through selection of candidates on phenotypic characteristics that are highly correlated with fillet traits. Studies have evaluated the ability of total weight and various body shape measurements to predict fillet traits and found body weight to be an accurate predictor of fillet weight (Rutten et al. 2004; Kause et al. 2007; Nguyen et al. 2010; Gjerde et al. 2012; Thodesen et al. 2012). By contrast, most studies obtained modest to low prediction accuracies for fillet percentage models developed from body measurements, although a few have shown potentially useful prediction power (Bosworth et al. 2001; Sang et al. 2009).

Prediction models for fillet yields are typically developed using relationships between growth metrics and fillet yields. There is a lack of information regarding phenotypic correlations between growth traits and processing characteristics for Asian sea bass stocks in Thailand. Therefore, the objective of this study was to investigate the potential for total weight and body measurements of live fish to predict fillet weights and fillet yields in four different stocks of Asian sea bass and to develop prediction equations for these traits.

Materials and methods

Fish origin and fry production

Sea bass fingerlings at approximately 90 days post-hatch (dph) were obtained from private hatcheries, including Chachoengsao (CHAC), Chon Buri (CHON), Chumphon (CHUM), and Samut Songkram (SAMU). The CHAC and SAMU stocks originated from wild populations of the Bang Pakong and Meklong Rivers, respectively, in the upper Gulf of Thailand. The CHON stock was founded from broodstock obtained from cage culture operations along the east coast of the Gulf of Thailand, mainly in Chon Buri Province. The CHUM stock originated from wild populations of the Chumphon River, which runs through the west coast of the Gulf of Thailand. These hatchery populations represent primary sources of seed stocks for Asian sea bass farming in Thailand (Department of Fisheries 2016; Senanan et al. 2015). In general, the hatcheries produced sea bass fry from group mating with a sex ratio of 1:2 (male:female). About 18–24 brood fish, aged 3–5 years with body weight of 4–5 kg were stocked in each spawning tank ($2 \times 4 \times 2.5 \text{ m}^3$). They were injected with 1 ml of synthetic hormone (Suprefact) under the dorsal fin, in the afternoon, to induce gamete maturation. Spawning normally took place 24–28 h post-injection (second night). Eggs were collected at 38–40 h (second morning) post-spawning and transferred to nursing tanks ($1.5 \times 2.5 \times 1 \text{ m}^3$), where larvae were reared until 20 dph. Fry were then transferred to concrete tanks ($2 \times 4 \times 1 \text{ m}^3$) and size graded at 30 and 60 dph at the hatcheries to minimize cannibalism.

Fish culture

The grow-out experiment was conducted at the Kasetsart University Fisheries Research Station in Samut Songkram Province. Four 1 m deep earthen ponds ($20 \text{ m} \times 40 \text{ m}$ or 800 m^2) were each equipped with four, 36 m^2 stationary cages ($6 \times 6 \times 1.2 \text{ m}$), providing a randomized complete block design (RCBD). In each pond (block), populations were randomly assigned to cages, each with 50 fish (a density of 1.4 fish per m²). Average size at stocking ranged from 9.48 ± 0.62 to 9.74 ± 0.16 g and 8.13 ± 0.82 to 9.04 ± 0.90 cm total



lengths. The cages were covered with netting to exclude predatory birds. Aeration was provided through polyethylene pipes on the pond bottoms below the cages to maintain sufficient dissolved oxygen concentration. Water was supplied from a brackish water canal connected to the Meklong River. Fresh well water was used to dilute brackish water to obtain salinities within 10–25 ppt. There was no water exchange during growout, but the ponds were topped off regularly to compensate for evaporation, maintaining the water depth within the cages at 80–90 cm. The cage netting was cleaned regularly to allow water flow through the cages.

Fish were fed to apparent satiation with a 40% protein commercial floating feed (Betagro Feed Mills, Bangkok, Thailand) twice daily at 07:00 and 17:00 h. Feed consumption was closely monitored and uneaten was removed prior to the next feeding. Fish health was observed and mortalities were immediately removed during feedings. Temperature and other water quality parameters were checked twice daily. Dissolved oxygen (DO) was measured using a portable model YSI 550A meter (Yellow Springs Instruments, Inc.; Yellow Springs, OH, USA) and pH using a model YSI EcoSense pH 100A meter (Yellow Springs Instruments, Inc.; Yellow Springs, OH, USA). Salinity was measured weekly using a portable salinity refractometer (Atago, Co., Ltd.; Tokyo, Japan). Aquatic plants and algae that may interfere with fish growth were removed regularly from the ponds.

After 300 days of rearing, fish were harvested and transported to the Taiban Fishery processing plant in Samut Prakarn Province, approximately 70 km from the research station. Fish were weighed (W in grams) and tagged to identify stock. They were placed on a light-colored background on their left side and photographed using a Nikon D40X digital camera. A centimeter scale was included in the images to ensure calibration of body measurement to the nearest 0.01 cm. SigmaScan Pro 4.0 software (www.sigmaplot.com) was used to measure 12 distances between body locations as illustrated in Fig. 1. A total of 400 fish (100 from each stock) were filleted by four skilled processors (P) at the processing plant, with each processor filleting individuals from all four stocks. The filleting process was as follows: (1) scales were removed manually. (2) Fish were cut from the top of the head, down the side behind the pectoral fins and along the side of the dorsal fins, and (3) the fillet was removed by cutting along the backbone from head to tail. The weight of the two fillets with skin attached (FW) was measured and recorded in grams. Fillet yield (FY) was calculated as $FY = (FW/W) \times 100$.

Statistical analysis

The distribution of data was examined using SAS PROC UNIVARIATE procedure (SAS Inst. Inc., Cary, NC, USA). The output indicated that all the analysis variables were normally distributed. Analysis of variance was performed to examine the effects of pond (block) and genetic factors (stock) on weight, standard length and

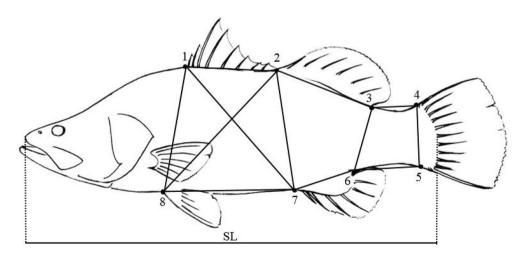


Fig. 1 Body measurement in Asian sea bass: distances from anterior 1st dorsal-anterior 2nd dorsal (M12), anterior 1st dorsal-anterior anal (M17), anterior 1st dorsal-anterior 2nd dorsal-posterior dorsal (M23), anterior 2nd dorsal-anterior anal (M27), anterior 2nd dorsal-anterior dorsal-dorsal membrane of caudal fin (M34), posterior dorsal-posterior anal (M36), dorsal-ventral membrane from caudal fin (M45), ventral membrane of caudal fin-posterior anal (M56), posterior anal-anterior anal-anterior pelvic (M78), and SL standard length



condition factor of preprocessed fish. There was a significant genetic effect on fish weights (P < 0.0001), whereas a block effect was not significant. Consequently, fillet yield was analyzed with total weight fitted as a covariate using the general linear model procedure (PROC GLM: SAS). The fixed effects of processor and stock were added to the model as: $y_{ijk} = \mu + P_i + S_j + \beta * W_{ijk} + e_{ijk}$, where y_{ijk} is an observation of FY, μ is the overall mean, P_i is a fixed effect of processor, S_j is a fixed effect of stock, β is the regression coefficient of body weight, W_{ijk} is a co-variable of total weight and e_{ijk} is a random residual error. In addition, fillet weight was analyzed with the fixed effects of processor and stock as: $y_{ijk} = \mu + P_i + S_j + e_{ijk}$, where y_{ijk} is an observation of FW and other parameters are as described previously.

Scatter plots revealed relationships between total weight and fillet traits (Fig. 2). Data for body weight and body measurements were pooled across stocks and phenotypic correlations among traits were calculated using Pearson correlations with PROC CORR in SAS. Residual correlations were calculated after adjustments for systematic effects (Bosworth et al. 2001; Hung and Nguyen 2014). Collinearity of independent variables was tested using a variance inflation factor (VIF). Due to high correlations among predictors, fillet weight was predicted from total weight and length using simple linear regression models (PROG REG) in SAS. Because there were many independent variables to consider, stepwise regression was used to develop prediction equations for fillet yields. Goodness of fit for the models was assessed using coefficient of determination (R^2) . In addition, the best model was determined based on conceptual predictive criterion (Cp) and Akaike's information criterion (AIC). Probability values < 0.05 were considered significant for estimates of parameters (intercept and beta). To assess the ability of these equations to predict new data, fillet records of each stock were evaluated using a fourfold cross-validation procedure. In the cross-validation, the data were randomly split into four subsets. Each iteration of the cross-validation used 75% of the records for training and the remaining 25% for testing. Predictive abilities of the equations were evaluated using the root mean squared error of prediction (RMSEP), calculated as the standard deviation of the residuals and the correlation between the predicted and the actual values (r) calculated for the best model. Prediction bias was calculated as the difference between the actual and the predicted values.

Results

Analyses of variance demonstrated highly significant differences in total weight and standard length among population samples of Asian sea bass (P < 0.0001) (Table 1). At 390 days of age, CHAC exhibited the highest weight (1166.32 \pm 23.42 g). Least squares means for total weight among the remaining three stocks were similar and ranged from 997.44 \pm 24.71 to 982.96 \pm 25.07 g. Analysis of processing yields

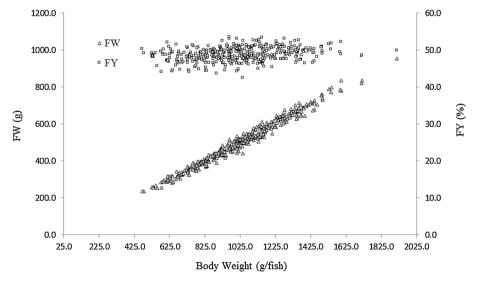


Fig. 2 Scatter plots showing relationships between body weight and fillet traits (FW fillet weight, FY fillet yield)



demonstrated significant differences among stocks. The CHAC samples exhibited the highest fillet weight of 596.23 ± 12.27 g, whereas weights among the remaining stocks were similar, ranging from 478.85 ± 13.32 to 491.53 ± 12.95 g. In addition, fillet yields differed among stocks with CHON ($49.88 \pm 0.19\%$) having the highest yield, followed by SAMU (48.24 \pm 0.19%), CHUM (47.80 \pm 0.20%), and CHAC (47.33 \pm 0.19%).

Phenotypic correlations of total weight with body measurements and fillet traits are presented in Table 2. Positive high correlations were observed for weight, standard length, and body measurements with fillet weight (0.99, 0.93, and 0.40-0.91). Significant but weak correlations were observed for total weight, standard length and body measurements with fillet yield (0.21, 0.24, and 0.11–0.24). A modest relationship (0.34) was observed between fillet weight and fillet yield.

Regression equations for predicting fillet weight from total weight, standard length, and body measurement (M18) in each stock are presented in Table 3. Across populations, total weight accounted for 98% of the variation in fillet weight, whereas standard length and M18 accounted for 86–91 and 61–84% of the variability in fillet weight, respectively. The model predictive power (RMSEP) for fillet weight indicated that the regression models developed from total weight resulted in a prediction error of 14.84–21.86 g. The errors of prediction were higher using standard length (39.84-44.99 g) and M18 (51.60-71.58 g) as predictors. Correlation coefficients between actual and predicted fillet weights were high for total weight (0.99). Fillet weights were under-predicted by 10.6, 0.38, and 6.03 g for CHAC, CHON and CHUM stocks, and overpredicted by 0.42 g for SAMU stock. Results for fillet yield predictions are presented in Table 4. Table 5 provides validation statistics for model selection. The best models were chosen when the Cp approached the number of parameters and AIC was smallest. The R^2 values were low for all predictors, which explained only 15-28% of the variation in fillet yield. The RMSEP ranged from 1.21 to 1.70%. Correlation coefficients between actual and predicted fillet yields were low for CHUM (0.20) and moderate for CHAC (0.49), CHON (0.51), and SAMU (0.39) stocks. Fillet yields were over-predicted by 0.03% for CHAC and under-predicted by 0.06, 0.04 and 0.06% for CHON, CHUM, and SAMU stocks, respectively.

Discussion

Results of the grow-out experiment indicated significant stock effects for all body size parameters. In particular, fillet weight varied significantly from 478.8 ± 13.32 g to 569.2 ± 12.27 g over the size range (960-1190 g) of fish examined. The CHAC stock exhibited greater total and fillet weights than the other stocks, but had the lowest fillet yield (47.3%). Interestingly, three stocks (CHON, CHUM and SAMU) exhibited similar weight were found to differ in fillet yields, among which CHON had the highest fillet percentage (49.88%). Such differences could result from variations in body shape among populations likely due to genetic composition and effects of selection within stocks. Differences in fillet yields (31–38%) between strains of Nile tilapia have been reported (Rutten et al. 2004; Peterman and Phelps 2012). Fillet

Table 1 Least squares means (± SE) for total weight (g), standard length (cm), fillet weight (g), and fillet yield (%) for Asian sea bass, after 300 days of rearing

Stock	Weight (g)	Length (cm)	Fillet weight (g)	Fillet yield (%)
CHAC	1166.32 ± 23.42^{a}	37.19 ± 0.25^{a}	569.23 ± 12.27^{a}	47.33 ± 0.19^{c}
CHON	982.96 ± 25.07^{b}	35.48 ± 0.27^{b}	486.10 ± 13.13^{b}	49.88 ± 0.19^{a}
CHUM	984.72 ± 25.42^{b}	35.37 ± 0.28^{b}	478.85 ± 13.32^{b}	$47.80 \pm 0.20^{\circ}$
SAMU	997.44 ± 24.71^{b}	35.83 ± 0.27^{b}	491.53 ± 12.95^{b}	48.24 ± 0.19^{b}
ANOVA				
F (stock/cage)	14.09	10.36	11.36	6.00
P value	< 0.0001	< 0.0001	< 0.0001	0.0005
F (processor)			0.46	0.86
P value			0.7127	0.4634

ANOVA results showing F values for differences among populations for preprocessed fish and among populations and processors for fillet weight and fillet yield. Entries in columns with different superscripts differ significantly at P < 0.05



Table 2 Pearson's correlation coefficients (above the diagonal) and residual correlations between total weight (W), standard length (L), body measurements M12, M17, M18,

M23, M27, M28, M34, M36, M45, M56, M67 and M78, fillet weight (FW) and fillet yield (FY)

Variables	W	L	M12	M17	M18	M23	M27	M28	M34	M36	M45	M56	M67	M78	FW	FY
W		0.93	0.77	0.91	0.87	0.81	0.87	0.90	0.52	0.84	0.76	0.40	0.52	0.84	0.99	0.21
L	0.92		0.76	0.91	0.86	0.76	0.88	0.89	0.53	0.83	0.74	0.41	0.48	0.86	0.93	0.24
M12	0.77	0.76		0.85	0.62	0.56	0.71	0.78	0.40	0.69	0.53	0.34	0.40	0.72	0.76	0.14
M17	0.92	0.91	0.86		0.85	0.79	0.91	0.95	0.47	0.87	0.73	0.34	0.44	0.92	0.91	0.22
M18	0.86	0.77	0.62	0.85		0.79	0.90	0.91	0.53	0.83	0.79	0.38	0.55	0.83	0.86	0.22
M23	0.81	0.77	0.56	0.79	0.78		0.80	0.78	0.38	0.75	0.75	0.22	0.63	0.76	0.80	0.19
M27	0.89	0.89	0.72	0.91	0.90	0.80		0.93	0.49	0.86	0.76	0.31	0.53	0.87	0.87	0.23
M28	0.91	0.90	0.79	0.95	0.91	0.77	0.93		0.52	0.89	0.78	0.34	0.50	0.90	0.89	0.24
M34	0.53	0.53	0.40	0.49	0.55	0.40	0.51	0.55		0.63	0.57	0.71	0.32	0.43	0.51	0.12
M36	0.85	0.83	0.68	0.87	0.84	0.75	0.87	0.88	0.64		0.75	0.42	0.50	0.80	0.83	0.21
M45	0.76	0.73	0.53	0.72	0.77	0.72	0.75	0.78	0.60	0.75		0.44	0.54	0.71	0.76	0.18
M56	0.29	0.39	0.33	0.36	0.39	0.24	0.33	0.38	0.71	0.43	0.49		0.15	0.25	0.40	0.11
M67	0.52	0.47	0.39	0.44	0.54	0.64	0.54	0.50	0.31	0.50	0.53	0.12		0.31	0.51	0.16
M78	0.87	0.88	0.75	0.36	0.39	0.75	0.33	0.39	0.46	0.43	0.70	0.31	0.32		0.84	0.23
FW	0.98	0.92	0.75	0.91	0.86	0.80	0.88	0.91	0.52	0.84	0.75	0.39	0.50	0.80		0.34
FY	0.25	0.27	0.16	0.23	0.25	0.21	0.23	0.25	0.13	0.21	0.20	0.14	0.17	0.22	0.39	

Table 3 Linear regression models showing intercept, regression coefficients, and coefficient of determination (R^2) for predicting fillet weight from total weight (W), standard length (L), and body measurement M18 in Asian sea bass stocks

Stock	Intercept	Regression coefficients			Prediction	P value	Validation		
		\overline{W}	L	M18	R^2		RMSEP	r	Bias
CHAC	- 37.70	0.52	_	-	0.97	< 0.0001	21.86	0.98	10.60
	- 1293.06	-	50.08	_	0.86	< 0.0001	44.99		
	- 584.01	_	_	98.84	0.61	< 0.0001	76.58		
CHON	- 31.14	0.53	_	_	0.98	< 0.0001	17.74	0.99	0.38
	- 1155.57	_	46.26	-	0.91	< 0.0001	39.80		
	- 708.32	_	-	107.16	0.84	< 0.0001	52.88		
CHUM	- 13.09	0.50	-	-	0.98	< 0.0001	18.04	0.99	6.03
	- 1036.36	_	42.84	_	0.90	< 0.0001	39.84		
	- 664.29	_	_	106.55	0.78	< 0.0001	59.71		
SAMU	- 26.69	0.52	_	_	0.98	< 0.0001	14.74	0.99	- 0.42
	- 1044.18	_	42.86	_	0.88	< 0.0001	40.29		
	- 726.49	_	-	108.61	0.80	< 0.0001	51.60		

Validation of prediction equations is based on root mean squared error of prediction (RMSEP), correlation coefficient between predicted and observed values (r), and prediction bias

percentages of Asian sea bass (47.3–49.8%) were similar to fillet yields observed in hybrid striped bass (49.7%) (Bosworth et al. 1998). This is not surprising as both species are anadromous perciformes fish, having slightly compressed, streamlined body shapes. In other species, fillet percentages varied from 35% for river catfish *Pangasianodon hypophthalmus* (Sang et al. 2009) and gilthead seabream *Sparus auratus* (Navarro et al. 2009) to 32–41% for common carp (Kocour et al. 2007; Bauer and Schlott 2009) and 42% for channel catfish (Argue et al. 2003). Fillet yields were as high as 58% for Coho salmon *Oncorhynchus kisutch* (Neira et al. 2004), 64% for rainbow trout *O. mykiss* (Kause et al. 2002), and 69% for Atlantic salmon *Salmo salar* (Powell et al. 2008). In giant freshwater prawn *Macrobrachium rosenbergii*, a meat yield of 45% was reported (Hung and Nguyen 2014). With regard to phenotypic correlations among growth metrics, fillet weight was closely



Table 4 Stepwise regression equations for predicting fillet yield from body measurements in Asian sea bass stocks. Predictor variables are standard length (L) and body measurements (M17, M18, M28, and M67)

Stock	Intercept	Regression coefficients							
		\overline{L}	M17	M18	M28	M67			
СНАС	30.88	0.67	- 0.34	- 0.06	_	_			
CHON	38.44	0.76	- 0.90	0.24	_	- 0.76			
CHUM	52.27	0.80	- 0.35	- 0.07	_	_			
SAMU	41.59	0.05	_	- 0.01	0.36	-			

Table 5 Validation statistics (conceptual predictive criterion, Cp and Akaike's Information Criterion, AIC), model property (coefficient of determination, R^2 , and root mean square of

prediction, RMSEP), and relationship between actual and predicted values (r) and prediction bias of the best models for predicting fillet yield in Asian sea bass populations

Population/traits	Ср	AIC	R^2	RMSEP	r	Bias	P value
CHAC							
L	0.0005	407.55	0.20	1.85			< 0.0001
L + M18	1.0026	405.36	0.22	1.84			0.0137
L + M18 + M17	2.0000	403.58	0.28	1.21	0.49	- 0.03	0.0102
CHON							
L	0.0021	349.98	0.15	1.78			< 0.0001
L + M18	1.0001	348.05	0.17	1.77			0.0003
L + M18 + M17	2.0021	345.85	0.20	1.73			0.0153
L + M18 + M17 + M67	3.0001	344.38	0.22	1.70	0.51	0.06	0.0096
CHUM							
L	0.0026	304.71	0.12	1.46			< 0.0001
L + M18	1.0032	303.26	0.14	1.45			0.0935
L + M18 + M17	2.0000	302.43	0.15	1.43	0.20	0.04	0.0002
SAMU							
L	0.0034	332.54	0.12	1.53			< 0.0001
L + M18	1.0000	330.71	0.15	1.52			0.0296
L + M18 + M28	2.0030	329.06	0.17	1.52	0.39	0.06	0.0002

related to fish weight, as well as standard length and body measurements, while there were weak correlations between fillet yield and fish size. Results were mostly in agreement with other previous studies, in that high correlations between fillet weight and total weight suggest that growing fish to a larger size is an efficient way to increase fillet weight.

Prediction equations developed in this study were based on simple linear regression, as some of the independent variables are highly correlated, e.g., total weight and standard length. In general, the most useful prediction equation is the one that accounts for the largest proportion of the variation (R^2) in the predicted traits. An R^2 of 1 would be a perfect prediction while an R^2 of 0 would indicate no predictive power. The predictions of fillet weight from total weight gave R^2 values of 0.98–0.99 and would be the equations recommended for use in practice for sea bass populations in this study. The proportion of variation explained by the prediction equations in our study using fish weight for fillet weight (0.98) was similar to the 0.95 R^2 reported for tilapia by Rutten et al. (2004). Similarly, Sang et al. (2009) predicted river catfish fillet weight of from body weight, and obtained an R^2 value of 0.86, which was lower than the values obtained in this study. However, a lesser degree of fit was exhibited when standard length (R^2 values of 0.86–0.81) or body measurement (M18) (R^2 values of 0.61–0.80) was used as predictors for fillet weight. In addition to R^2 values, RMSEP and correlation coefficients between observed and predicted values (r) of the equations should be considered if predictions are to be made for Asian sea bass stocks other than those in this study. Validation of the model developed from body weight showed the highest correlation coefficients between predicted and



observed fillet weights (0.98–0.99) and lowest RMSEP values (14.74–21.86 g), suggesting that these equations would be applicable to other Asian sea bass stocks.

Fillet yield, as a proportional trait consisting of fillet weight and total weight, showed weak relationships to total weight, standard length, and body measurements. Total weight is an important variable, normally included in prediction equations for fish fillet yield. The proportion of variation in fillet yield explained by the regression equation was low to moderate ($R^2 = 0.16-0.36$) for total weight in tilapia (Rutten et al. 2004), but much higher for the ratio between body volume and body weight in river catfish (0.75) (Sang et al. 2009). However, only body measurement traits not total weight were included in the stepwise-developed regression equations in our study. The first included variable was standard length, which accounted for 12-20% of the variation in fillet yields, depending on the stock. Addition of other variables (M17, M18, M28 and M67) depending on the stock accounted for an additional 5-8% of the variation in fillet yields. Although the best model for each population gave a low to moderate degree of predictive ability for fillet yield, prediction bias was close to zero. The relationship between fish size and fillet yield in this study generally confirms the principle that larger fish provide higher fillet yield percentages. However, regarding the prediction equations, for example, the regression coefficient of standard length on fillet yield was 0.67 for the CHAC stock. This value represents the expected change in the predicted fillet yield for each unit change in fish length, when the remaining variables are fixed. Therefore, for each 1 cm increase in body length, an individual fillet yield is expected to increase 0.67%. This prediction ability for predict fillet yield is probably too low to be practical for predicting new data for the CHUM stock (r = 0.20). At this stage, it appears that our regression models for predicting fillet yields in three stocks are adequate (r values of 0.39–0.51). For future work, it would be useful to develop similar models for larger fish (1500–2000 g) using a larger data set to increase predictive power. With the use of up to 2700 records and a multiple regression models, Sang et al. (2009) obtained an r value of 0.86 for predicting fillet yield in river catfish.

In conclusion, total weight is the best single measurement for predicting fillet weight in Asian sea bass. Regression models developed in this study are applicable to fish weighing 960–1190 g from all populations sampled. Using separate models yielded similar prediction power and accuracy for fillet weight. This study demonstrates the use of various body measurements in predicting fillet yields. Despite relatively weak correlations between body measurements and fillet yields, we were able to develop useful prediction models for fillet yields in three stocks of Asian sea bass. Although fillet yield is the most important trait for fish processors, increased fillet percentages could be achieved along with increased fillet weights, given the positive correlation between the traits. The application of prediction models would allow for rapid genetic improvement via breeding to increase fillet yields and enable producers to be more competitive in the value-added markets of Asian sea bass.

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