

1 **Novel epigenetic age estimation in wild-caught Gulf of Mexico reef fishes**

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3 D. Nick Weber^{1*}, Andrew T. Fields¹, William F. Patterson III², Beverly K. Barnett^{2,3},

4 Christopher M. Hollenbeck¹, and David S. Portnoy¹

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6 ¹ Texas A&M University–Corpus Christi, Department of Life Sciences, 6300 Ocean Drive,

7 Corpus Christi, TX 78412, USA.

8 ² University of Florida, School of Forest, Fisheries, and Geomatics Sciences, 7922 NW 71st

9 Street, Gainesville, FL 32653, USA.

10 ³ NOAA Fisheries, Southeast Fisheries Science Center, Panama City Laboratory, 3500 Delwood

11 Beach Road, Panama City, FL 32408, USA.

12

13 ***Corresponding author:** D. Nick Weber (email: dweber@islander.tamucc.edu)

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15 **Competing interests:** The authors declare there are no competing interests.

16 **Abstract:** Cutting-edge DNA methylation-based epigenetic aging techniques were applied to
17 Gulf of Mexico northern red snapper (*Lutjanus campechanus*; $n = 10$; 1–26 years old) and red
18 grouper (*Epinephelus morio*; $n = 10$; 2–14 years old). Bisulfite-converted restriction site-
19 associated DNA sequencing was used to identify CpG sites (cytosines followed by guanines) that
20 exhibit age-correlated DNA methylation, and species-specific epigenetic clocks developed from
21 100s of CpG sites in each species showed strong agreements between predicted and otolith-
22 derived ages ($r^2 > 0.99$ for both species). Results suggest epigenetic age estimation could provide
23 an accurate and efficient approach to mass-aging fishes in a non-invasive manner.

24

25 **Keywords:** DNA methylation, epigenetic clock, age estimation, stock assessment, fisheries, red
26 snapper, red grouper

27 **Introduction**

28 Age data are fundamental to determining life history parameters (e.g. age-at-length, age-
29 at-maturity, age-related fecundity) and are thus critical for fisheries assessment and management,
30 particularly when age-structured stock assessment models are used to estimate if a stock is
31 overfished or undergoing overfishing. Fish age is traditionally determined by counting growth
32 zones in a range of hard structures, including otoliths, vertebrae, scales, and fin rays (Campana
33 2001). Such techniques can be costly and time intensive (Helsler et al. 2019), are of low accuracy
34 for some species, are subject to reader bias, and are necessarily lethal in the case of otoliths and
35 vertebrae (Campana 2001; Anastasiadi and Piferrer 2019). Moreover, protected species, brood
36 stock for hatchery programs, and commercially valuable whole fish cannot have their otoliths or
37 vertebrae extracted, and thus cannot be accurately aged using traditional techniques. As the
38 demand for fish age composition data is increasing (Helsler et al. 2019), there is a need to
39 develop alternative age estimation methods.

40 Epigenetics refers to molecular-level mechanisms that affect gene expression without
41 altering the underlying DNA sequence and that are heritable down cell lines or from parent to
42 offspring (Kilvitis et al. 2014). DNA methylation is the most studied epigenetic mechanism and
43 refers primarily to the addition of methyl groups (CH₃) to cytosines located within CpG
44 dinucleotides (cytosines followed by guanines; Kilvitis et al. 2014). Recent studies have
45 demonstrated that changes in DNA methylation levels at certain CpG sites exhibit strong
46 correlations with chronological age, leading to the development of age-predictive models based
47 on DNA methylation, referred to as epigenetic clocks (reviewed in Parrott and Bertucci 2019).
48 Previously developed epigenetic clocks have involved methylation levels at 3 to 353 CpG sites
49 (Horvath 2013; Polanowski et al. 2014) identified in a range of tissue types, including skin

50 (Polanowski et al. 2014), blood (Thompson et al. 2018), muscle (Anastasiadi and Piferrer 2019),
51 and fin clips (Mayne et al. 2020), and patterns have been shown to be tissue-specific (Horvath
52 2013; Thompson et al. 2018). While epigenetic clocks have been derived predominantly for
53 mammalian species, a handful of epigenetic clocks have been developed for laboratory-raised
54 fishes, including for European sea bass (*Dicentrarchus labrax*; Anastasiadi and Piferrer 2019),
55 zebrafish (*Danio rerio*; Mayne et al. 2020), and medaka (*Oryzias latipes*; Bertucci et al. 2021).
56 In addition, Mayne et al. (2021) developed epigenetic clocks for three species of threatened
57 fishes (Australian lungfish, *Neoceratodus forsteri*; Murray cod, *Maccullochella peelii*; and Mary
58 River cod, *Maccullochella mariensis*), using a combination of wild and laboratory-raised
59 individuals. While the development of epigenetic clocks in the aforementioned species suggests
60 DNA methylation levels may generally be predictive of age in fishes, the majority of the
61 aforementioned clocks were developed using samples from fishes reared in laboratories. Fish
62 reared in controlled environments are not exposed to the same degree of environmental variation
63 experienced by fishes in the wild and thus may not exhibit the same degree of environmentally
64 induced epigenetic change. Moreover, Mayne et al. (2021) targeted CpG sites with methylation
65 levels that are known to significantly correlate with age in laboratory-raised zebrafish and that
66 were conserved between zebrafish and the target species. Therefore, the potential *de novo*
67 identification of CpG sites exhibiting age-correlated methylation in wild-caught fishes of
68 management concern, and the use of such sites to accurately estimate age, remains unknown.

69 Red snapper (*Lutjanus campechanus*) and red grouper (*Epinephelus morio*) are two of the
70 most important species in commercial and recreational fisheries in the U.S. Atlantic and Gulf of
71 Mexico. Both species are long-lived (>50 years for red snapper and >25 years for red grouper),
72 and both are of management concern (SEDAR 52; SEDAR 61). Age estimates for both species

73 have been generated through traditional otolith increment analysis (Lombardi-Carlson et al.
74 2002; Lombardi 2017), and otolith-derived age estimation for red snapper has been validated
75 with the bomb radiocarbon chronometer (Barnett et al. 2018). However, as with many managed
76 species, the regulatory histories of red snapper and red grouper stocks have been complex,
77 requiring large investments of time, effort, and financial resources into processing ever-
78 increasing numbers of samples to generate age composition data that is input into stock
79 assessment models (Passerotti et al. 2020). For example, over 49,000 otolith samples were
80 processed and aged for the most recent stock assessment of Gulf of Mexico red snapper
81 (Lombardi 2017).

82 Given the importance of age data for fisheries assessments and the lack of effective non-
83 lethal methods for estimating age, the development of epigenetic clocks in species of
84 management or conservation concern would be of great utility. In the present study, we sought to
85 investigate the possibility of creating epigenetic clocks for wild-caught red snapper and red
86 grouper, through the *de novo* identification of loci exhibiting age-correlated DNA methylation in
87 two different tissue types (muscle and fin clips), both of which can be sampled non-lethally in
88 the course of fisheries research.

89

90 **Methods**

91 Tissues were obtained from red snapper (muscle, $n = 10$ individuals) and red grouper (fin
92 clips, $n = 10$ individuals) collected in the northern Gulf of Mexico and for which age was
93 estimated via counts of opaque zones in otolith thin sections (range: 1–26 years for red snapper
94 and 2–14 years for red grouper). In addition, all red snapper samples analyzed in this study had
95 their ages directly validated via application of the bomb radiocarbon chronometer (Barnett et al.

96 2018; Patterson et al. 2021). Genomic DNA was extracted from tissue samples using a Mag-
97 Bind Blood & Tissue DNA Kit (Omega Bio-tek, Inc., Norcross, USA), and DNA libraries were
98 prepared for bisulfite-converted restriction site-associated DNA sequencing (bsRADseq),
99 following a modified version of the Trucchi et al. (2016) protocol (described in Supplemental
100 Material¹). Libraries for both species were split into two portions, and one portion was bisulfite-
101 treated using an EpiTect Plus Bisulfite Kit (Qiagen, Hilden, Germany). Bisulfite treatment
102 converts unmethylated cytosines into uracils through chemical deamination, and uracils are
103 subsequently replaced by thymines during PCR. This results in predictable base substitutions at
104 all unmethylated cytosines, which can be identified by comparing sequences from the treated
105 library to the untreated library (Trucchi et al. 2016). Libraries were sequenced across a single
106 lane per species on an Illumina HiSeq4000 (Illumina, Inc., San Diego, USA). Red snapper reads
107 were mapped to a draft of the red snapper genome (Portnoy, D.S. unpublished data), while the
108 red grouper bisulfite-treated reads were mapped to a reference genome constructed from the
109 untreated reads, using the *dDocent* pipeline (Puritz et al. 2014) for $c = 0.88$, $K_1 = 2$, and $K_2 = 1$.
110 Mapped reads were then filtered to retain primary alignments, proper pairs, and those with a
111 mapping quality ≥ 40 .

112 CpG sites that could not be successfully genotyped in the untreated library or that were
113 identified as potential single nucleotide polymorphisms (SNPs; defined as sites where $>5\%$ of
114 total untreated reads across individuals displayed a cytosine to thymine substitution on the
115 forward strand or guanine to adenine substitution on the reverse strand) were removed from the
116 dataset. For each species, CpG sites were then filtered to retain only sites present in $\geq 90\%$ of
117 individuals. To identify CpG sites exhibiting age-correlated methylation, a Bayesian framework

¹Supplementary data are available online.

118 was used to estimate the parameters of a generalized linear model (GLM) that included otolith-
119 derived age as a fixed factor and individual as a random factor, using the *rstanarm* package
120 (version 2.19.3; Goodrich et al. 2020; described in Supplemental Material¹). The response
121 variable was the binomial expression of the number of methylated reads (n) and the total number
122 of reads for each sample (k) at a given CpG site $\binom{n}{k}$. GLMs were considered to have converged if
123 the n_{eff} , a crude measure of the effective sample size, was greater than 2,000 or the Gelman-
124 Rubin convergence diagnostic was less than 1.01 (Lunn et al. 2013; Muth et al. 2018). CpG sites
125 with a 95% credible interval that did not include zero for the slope of otolith-derived age versus
126 methylation were considered to exhibit significant age-correlated methylation.

127 For each CpG site that exhibited age-correlated methylation, percent methylation was
128 estimated as the number of methylated reads divided by the total number of reads, and per-site
129 95% confidence intervals were calculated around the estimate in each individual (Clopper and
130 Pearson 1934). Only those sites with confidence intervals <0.60 in at least 8 individuals per
131 species were retained (Thompson et al. 2017). For retained sites, individuals with over-dispersed
132 confidence intervals (>0.60) were entered as missing data. Because downstream analysis does
133 not allow for missing data, methylation frequencies at missing individuals/sites were imputed
134 using the *MICE* package (version 3.13.0; van Buuren and Groothuis-Oudshoorn 2011) in *R*
135 (version 3.6.0).

136 The relationship between otolith-derived age and percent methylation across CpG sites
137 was characterized using elastic net penalized regression modeling, a method that reduces the
138 magnitude of coefficients and the number of predictor variables to reduce model complexity.
139 Modeling was implemented in *glmnet* (version 4.0.2; Friedman et al. 2010) with an alpha
140 parameter of 0.5 (considered the optimal merging of a ridge and lasso model; Thompson et al.

141 2017; Mayne et al. 2020; Bertucci et al. 2021). The internally cross-validated version of *glmnet*
142 (*cv.glmnet*) was utilized to automatically select the optimal penalty parameter (λ), with 10-fold
143 cross validation. Linear regressions were then computed to visualize the relationship between
144 predicted age and otolith-derived age.

145

146 **Results**

147 *Red snapper*

148 For red snapper, 1,674,121 CpG sites were recovered across all samples. A total of
149 408,745 sites either could not be successfully genotyped in the untreated library or were
150 identified as potential SNPs and were removed from the dataset. Of the remaining 1,265,376
151 sites, 49,189 sites were present in at least 90% of samples. Mean global CpG methylation across
152 the retained sites was 85.72% ($\pm 0.31\%$; SD). Bayesian GLMs identified 3,224 sites that
153 exhibited significant age-correlated methylation, and a total of 1,829 sites had sufficiently tight
154 confidence intervals in at least 80% of samples. Penalized regression analysis retained 199 CpG
155 sites in the age-predictive model ($\lambda = 0.1104$). The slope and y intercept (\pm SE) for the regression
156 of predicted age versus otolith-derived age were 0.98 (± 0.01) and 0.25 (± 0.09 ; $r^2 = 0.9995$; Fig.
157 1A). Of the 199 sites included in the final model, 80.40% of sites ($n = 160$) exhibited
158 hypomethylation (i.e., a decrease in methylation with increasing age; Fig. 1C).

159 *Red grouper*

160 For red grouper, 1,238,719 CpG sites were recovered across all samples. A total of
161 163,925 sites either could not be successfully genotyped in the untreated library or were
162 identified as potential SNPs and were removed from the dataset. Of the remaining 1,074,794
163 sites, 9,834 sites were present in at least 90% of samples. Mean global CpG methylation across

164 the retained sites was 85.45% ($\pm 1.08\%$; SD). Bayesian GLMs identified 690 sites that exhibited
165 significant age-correlated methylation, and a total of 307 sites had sufficiently tight confidence
166 intervals in at least 80% of samples. Penalized regression analysis retained 49 CpG sites in the
167 age-predictive model ($\lambda = 0.0501$). The slope and y intercept (\pm SE) for the regression of
168 predicted age versus otolith-derived age were 0.98 (± 0.01) and 0.17 (± 0.05 ; $r^2 = 0.9996$; Fig.
169 1B). Of the 49 sites included in the final model, 75.51% of sites ($n = 37$) exhibited
170 hypomethylation (Fig. 1D).

171

172 Discussion

173 The present study demonstrates the *de novo* identification of loci exhibiting age-
174 correlated DNA methylation in wild-caught fishes and the utility of using such loci to estimate
175 age. The epigenetic clocks developed show strong agreement between predicted and otolith-
176 derived ages ($r^2 > 0.99$ for both species), supporting the notion that this may be an important yet
177 currently untapped tool for fisheries science.

178 The epigenetic clocks developed for red snapper and red grouper performed well relative
179 to those previously developed (Anastasiadi and Piferrer 2019; Mayne et al. 2020, 2021; Bertucci
180 et al. 2021). The strong agreement between predicted and otolith-derived ages reported here is
181 promising and may be the result of characterizing a large number of independent CpG sites
182 exhibiting age-correlated methylation in both species. However, the low sample sizes ($n = 10$ for
183 both species) precluded the application of more rigorous predictive approaches, involving
184 training and testing datasets (Mayne et al. 2021) to minimize model overfitting. To more
185 thoroughly validate the predictive capability of these epigenetic clocks, larger numbers of
186 samples spread evenly across ages, with replicates at ages, will be necessary. In addition, while

187 the results reported here indicate that the epigenetic clocks developed are potentially highly
188 accurate, a larger number of samples is needed to assess precision.

189 While previously developed epigenetic clocks involved fishes reared in the laboratory
190 (Anastasiadi and Piferrer 2019; Mayne et al. 2020; Bertucci et al. 2021), the results of the present
191 study indicate accurate epigenetic clocks can also be developed for wild-caught fisheries species.
192 Changes in patterns of DNA methylation can be induced by aspects of the environment, such as
193 water temperature, salinity, and toxin levels (reviewed in Beal et al. 2018). Thus, fish reared
194 under controlled conditions will not experience the same degree of environmental fluctuation
195 (process error), and therefore environmentally induced epigenetic change, as fishes in the wild.
196 Having said that, the predictive capabilities of epigenetic clocks previously developed for
197 laboratory-raised European sea bass (Anastasiadi and Piferrer 2019) and medaka (Bertucci et al.
198 2021) were not significantly affected by increases in water temperature or exposure to ionizing
199 radiation, respectively. Nonetheless, the results presented here not only indicate that accurate
200 epigenetic clocks can be developed for wild-caught fishes despite potential noise introduced by
201 environmental heterogeneity, but also that different tissue types (muscle and fin clips), both of
202 which can be sampled non-lethally, are appropriate for such work. Because studies in humans
203 (Horvath 2013) and mice (Thompson et al. 2018) have shown that epigenetic clocks can be
204 tissue-specific, comparisons across tissue types within species will be important to consider
205 when creating epigenetic clocks for fisheries applications. Finally, hypomethylation was
206 observed in a greater number of sites in the final models for both species and among all sites
207 identified as exhibiting significant age-correlated methylation (58.86% and 74.44% for red
208 snapper and red grouper, respectively). This finding is consistent with other studies that have

209 observed higher frequencies of age-correlated hypomethylation than age-correlated
210 hypermethylation (reviewed in Johnson et al. 2012).

211 The ability to identify loci exhibiting age-correlated DNA methylation in wild-caught
212 fishes through reduced representation sequencing approaches, demonstrated herein, suggests
213 epigenetic clocks could be a widely applied tool for fisheries research. The development of
214 epigenetic clocks would allow for the collection and incorporation of age-specific biological data
215 into stock assessments without the need for destructive sampling, increasing the diversity of data
216 sources that could be accessed. Moreover, age data are currently considered to be one of the
217 more expensive sources of data utilized in stock assessments, due to the time required for otolith
218 processing and analysis (Helser et al. 2019). The development of epigenetic clocks could allow
219 for the construction of high-throughput multiplex PCR assays (Mayne et al. 2021) to assess
220 specific age-correlated CpG sites (50-500 sites; Campbell et al. 2015), thus enabling accurate,
221 rapid, cost-effective (<\$10 per sample; Mayne et al. 2021), and concurrent age estimation in
222 thousands of individuals.

223

224 **Funding statement**

225 This work was funded by a NMFS-Sea Grant Joint Fellowship in Population and
226 Ecosystem Dynamics (No. NA20OAR4170466) from the National Marine Fisheries Service and
227 the National Sea Grant College Program and a 'Grant-In-Aid' (No. NA18-E/GIA-2020-Weber)
228 from the Texas Sea Grant 'Grants-In-Aid' of Graduate Research Program. This is publication 29
229 of the Marine Genomics Laboratory and 122 of Genetic Studies in Fishes (Genetic Studies in
230 Marine Fishes).

231

232 **Contributors' statement**

233 **DNW:** data collection, methodology, formal analysis, writing (original draft); **ATF:**
234 conceptualization, methodology, formal analysis, writing (review and editing); **BKB:** data
235 collection, writing (review and editing); **WFP:** conceptualization, data collection, writing
236 (review and editing); **CH:** conceptualization, resources, writing (review and editing); **DSP:**
237 conceptualization, methodology, resources, writing (review and editing).

238

239 **Data availability statement**

240 Datasets (raw and filtered) and scripts (including those used to create a *de novo* reference
241 genome for red grouper, to conduct the Bayesian GLMs, to calculate 95% confidence intervals,
242 and to conduct the elastic net regressions) are available at
243 <https://github.com/marinegenomicslab/Epi-Age-Est>. Raw bsRADseq sequences will be made
244 publicly available at the conclusion of a separate ongoing study.

245

246 **References**

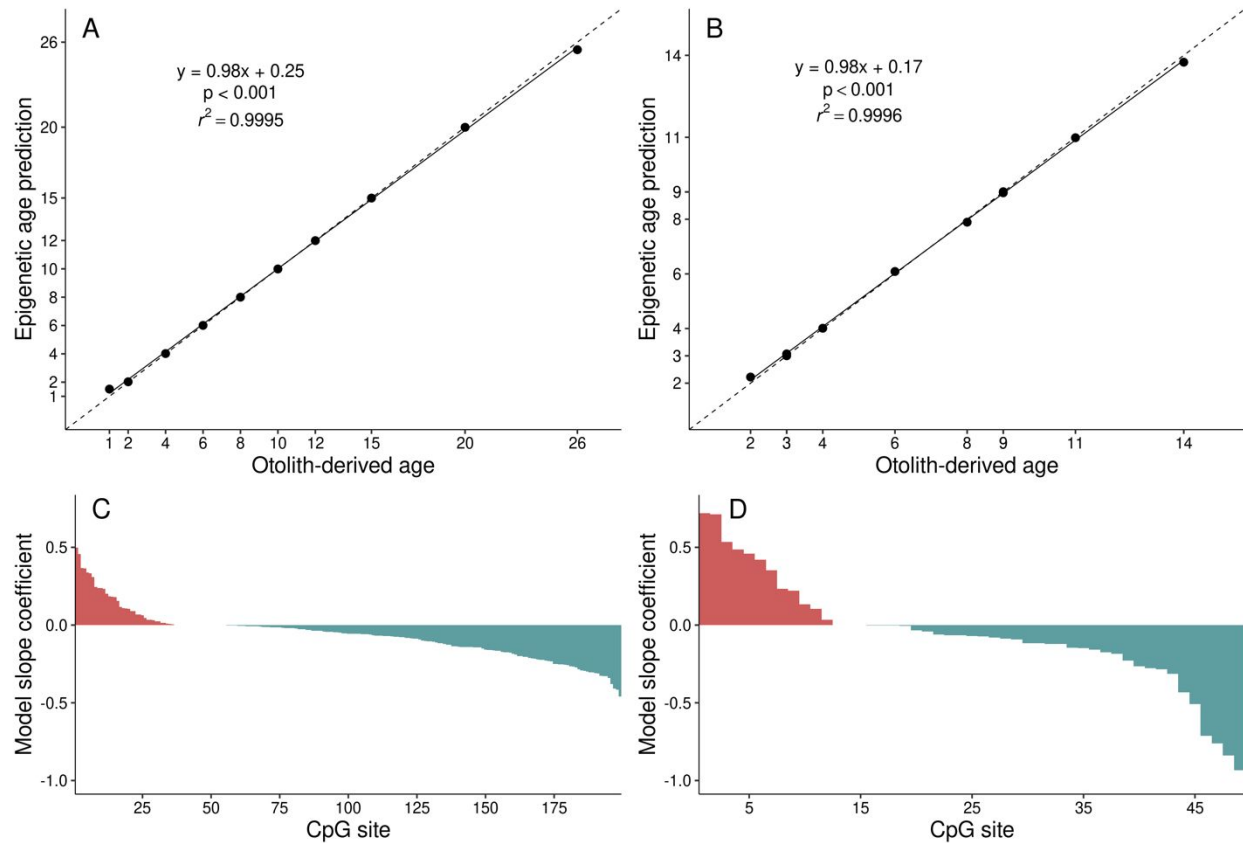
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326

327 **Fig. 1.** Epigenetic age predictions versus otolith-derived ages for (A) red snapper ($n = 10$; 1–26
 328 years old) and (B) red grouper ($n = 10$; 2–14 years old). Dashed lines indicate lines of 1:1
 329 agreement between predicted and otolith-derived ages. Solid lines represent linear regression fits
 330 to the data, with equations indicated on panels. Panels C and D depict the final model slope
 331 coefficients for each CpG site included in the final model for red snapper and red grouper,
 332 respectively. Red bars indicate CpG sites exhibiting hypermethylation, while blue bars indicate
 333 CpG sites exhibiting hypomethylation.