

1 **Status and targets for rebuilding the three major fish stocks in**  
2 **Lake Victoria**

3 Laban Musinguzi\*, Mark Olokotum & Vianny Natugonza

4 National Fisheries Resources Research Institute, P.O. Box, 343,  
5 Jinja, Uganda

6 \*Corresponding author: [labanmusinguzi@firi.go.ug](mailto:labanmusinguzi@firi.go.ug)

7

8 **Abstract**

9 We determined fisheries management reference points for three major  
10 fish stocks in Lake Victoria (Nile tilapia, Nile perch and Dagaa)  
11 for Uganda and the whole lake. The aim was to ascertain stock status  
12 and define reasonable objectives and targets for rebuilding to  
13 sustainable levels. Dagaa was found to be healthy in Uganda and the  
14 whole lake but tending to overfished status. In Uganda, the stock  
15 status of Nile tilapia and Nile perch was recruitment impaired but  
16 tending more towards collapsed and overfished status respectively.  
17 In the whole lake, the stock status of Nile tilapia and Nile perch  
18 was collapsed and overfished respectively with the latter tending  
19 more towards recruitment impaired. Estimates of maximum sustainable  
20 yield (MSY) showed that catches could be increased under good  
21 management. Rebuilding the Nile tilapia and Nile perch stock  
22 biomasses to MSY level ( $B_{msy}$ ) could respectively increase the catches  
23 above the current level by 9.2% and 29.5% in Uganda and by 72.8% and  
24 15.1% in the whole lake. The immediate objective for fisheries  
25 management should be to rebuild biomass for the Nile tilapia and  
26 Nile perch stocks to  $B_{msy}$ . Elimination of illegal fishing practices  
27 has proved to be effective. In addition, management needs to keep  
28 catches at low levels until biomass for the stocks is  $\geq B_{msy}$  for at  
29 least three consecutive years.

30 **Introduction**

31 Uganda and Tanzania have since 2017 strengthened enforcement of  
32 fisheries regulations to end illegal fishing and improve stocks of  
33 Lake Victoria. With ~1 million tons of fish produced annually, Lake  
34 Victoria supports the world's biggest inland fishery useful for  
35 foreign exchange, employment and direct sustenance of >4 million  
36 people (Marshall & Mkumbo 2011). These benefits have for a long time  
37 been threatened by high fishing pressure (Njiru et al, 2007;  
38 Nyamweya et al. 2020), justifying the strengthened enforcement.

39

40 The countries strengthened the enforcement by deploying their  
41 respective defense forces, Fish Protection Unit (FPU) in Uganda and  
42 the multisector task force in Tanzania (Mudliar, 2018; NPA, 2019).  
43 These partially or fully replaced previous institutional  
44 arrangements such as beach management units which were considered  
45 ineffective because of corruption (Nunan et al. 2018). In Uganda,  
46 the enforcement demonstrated determination to end illegal fishing  
47 practices as the deployment was followed by a total stop on all  
48 forms of illegalities. The ineffective institutional arrangements  
49 were replaced, illegal gears and crafts destroyed, fishers forced  
50 out of near shore areas.

51

52 Positive outcomes have been observed from the enforcement. Data from  
53 fishery independent surveys conducted since 2017 show that the  
54 biomass of Nile perch (*Lates niloticus* (Linnaeus, 1758)), the most  
55 important commercial fish species in the lake has improved and was  
56 at its record high since 2010 (Hydro-acoustics Regional Working  
57 Group, 2019). Interestingly, 48% of the increase was recorded  
58 between 2018 and 2019, with the largest increase recorded in the  
59 Ugandan part followed by Tanzania. The increase in biomass was least  
60 in Kenya where enforcement was not strengthened. The surveys further  
61 showed that although Nile perch was still dominated by individuals  
62 under the size at which recruitment to the fishery occurs, there was  
63 a record increase in the proportion of fish at the preferred size.  
64 These observations suggest that good management in Lake Victoria  
65 can pay off.

66

67 With all due respect, the enforcement is ongoing with no  
68 consideration of fisheries management reference points, lacking  
69 clear management objectives and targets beyond the elimination of  
70 the illegal fishing gears and practices. To contribute to effective  
71 enforcement, we estimated fisheries management reference points for  
72 major commercial fish species to act as a basis of adopting  
73 evidence-based fisheries management objectives and targets.  
74 The reference points determined for the whole lake and the Ugandan  
75 part of the lake clarify on the status of the stocks before the  
76 commencement of the enforcement. They are not only useful for  
77 substantiating objectives and targets for the enforcement but are  
78 also indispensable for evaluating its effectiveness. At the global  
79 scale, this assessment is commensurate with calls to increase  
80 assessment of inland fish stocks to support responsible inland  
81 fisheries (Cooke et al. 2016, FAO & MSU, 2015; FAO, 2020).

82

## 83 **Methods**

### 84 **Approach and stocks assessed**

85 The reference points were based on two methods, a Monte Carlo method  
86 (CMSY) and a Bayesian state-space implementation of the Schaefer  
87 production model (BSM). A brief background to these methods is  
88 provided here while more details are available in Froese et al.  
89 (2017). The methods are built on a principle that catch from a  
90 species is produced by its biomass and productivity such that if two  
91 of the three parameters are known, production models can be used to  
92 estimate the other. The CMSY uses catch and productivity to estimate  
93 biomass. The method uses prior ranges of productivity and current  
94 biomass (B) relative to unexploited biomass (k) (B/k) at the start,  
95 intermediate and end of a time series to detect productivity and  
96 unexploited biomass pairs with corresponding biomass estimates that  
97 are compatible with observed catches. The BSM on the other hand uses  
98 catch and biomass data to estimate productivity. The methods are  
99 integrated with other empirical formulae to estimate the reference  
100 points including maximum sustainable yield (MSY), fishing mortality  
101 rate F at MSY ( $F_{msy}$ ), biomass required to support MSY ( $B_{msy}$ ), relative  
102 stock size ( $B/B_{msy}$ ) and exploitation ( $F/F_{msy}$ ).

103

104 We estimated reference points for three major stocks in Lake  
105 Victoria at two spatial scales: the whole lake and the Ugandan part  
106 of the lake. Nile perch, *Oreochromis niloticus* (Linnaeus, 1758)  
107 (Nile tilapia) and *Rastrineobola argentea* (Pellegrin, 1904) locally  
108 known as Dagaa are the three major fish species supporting  
109 commercial fisheries in Lake Victoria. The species are responsible  
110 for >88.7% (estimated from catch used in this study) of catches from  
111 Lake Victoria. Dagaa is the most important by weight, followed by  
112 Nile perch and Nile tilapia. Nile perch supports fish processing  
113 industries that export to foreign markets including the European  
114 Union and is the most important by value.

115 **Data requirements, sources and application to CMSY**

116 To estimate the reference points, abundance and catch data were  
117 required at the two spatial scales. For the whole lake, we estimated  
118 the reference points using two indices of abundance i.e. absolute  
119 biomass and fishery independent catch per unit effort (CPUE). This  
120 provided an opportunity to evaluate the usefulness of both sets of  
121 data which are available for Lake Victoria. The reference points for  
122 the Ugandan part of the lake were based on CPUE only.

123

124 The absolute biomass was obtained from Nyamweya et al. (2016) who  
125 simulated the biomass based on catches and hydrodynamics of the lake  
126 using ecosystem models. This was available to 2015, starting from  
127 1965, 1968 and 1971 for Nile perch, Dagaa and Nile tilapia  
128 respectively. The CPUE was from hydroacoustic (Nile perch and Dagaa)  
129 and trawl (Nile tilapia) surveys and was only consistent for the  
130 species since 1999. The CPUE was restricted to 2015 beyond which no  
131 catch data are available.

132

133 Catch data used for the whole lake was partly available from  
134 Nyamweya et al. (2016) and was supplemented with data from the  
135 archives of the National Fisheries Resources Research Institute.  
136 The archives were the sources of the catch data for the Ugandan  
137 part.

138

139 Productivity of a stock is reflected in CMSY as prior ranges of  
140 intrinsic rate of population increase ( $r$ ) which are derived by  
141 classifying resilience of species available in FishBase into  $r$   
142 values (Froese & Pauly, 2015; Froese & Pauly, 2019). The resilience  
143 of Dagaa is high and that of Nile tilapia and Nile perch is medium.  
144 Their respective  $r$  ranges are 0.6-1.5, 0.2-0.8 and 0.2-0.8 (Froese &  
145 Pauly, 2019; Froese et al. 2017).

146

147 The  $B/k$  prior ranges depend on depletion status of stocks: very  
148 strong depletion (0.01 - 0.2), strong depletion (0.01 - 0.4), medium  
149 depletion (0.2 - 0.6), low depletion (0.4 - 0.8), and nearly  
150 unexploited (0.75 - 1.0) (Froese et al. 2019). These are required  
151 for the start, intermediate and end year of the time series. We  
152 harnessed trends in biomass and catches over the time series to set  
153 the  $B/k$  ranges for the species (Figure 1).

154

155 The start years for the Nile perch, Nile tilapia and Dagaa were  
156 1965, 1971, 1968 respectively in the first scenario for the whole lake

157 using absolute biomass, and 1999 for the second whole lake scenario  
158 using CPUE and the Ugandan part. The end year was 2015 in both  
159 scenarios.

160

161 For the first whole lake scenario, the B/K ranges at the start years  
162 were set at low depletion for all the stocks (Table 1). In these  
163 years, the stocks were at the initiation fishery development phase  
164 and the low depletion enabled the future development of their  
165 fisheries (Hilborn & Walters, 1992; Figure 1). At the end, the Nile  
166 perch fishery had reached the decline phase of fishery development  
167 (Hilborn & Walters, 1992; Figure 1) which was characterized by high  
168 fishing effort (Nyamweya et al. 2020). For this reason, we set the  
169 B/k priors to strong depletion. For Nile tilapia, by 2015, although  
170 the catch in the whole lake was not the lowest ever, it was 10.3% of  
171 the historical maximum, guiding us to set the B/K prior to very  
172 strong depletion. For Dagaa, the catch was increasing albeit  
173 decreasing biomass. Given its high turnover rate and high fishing  
174 pressure (Mangeni-Sande et al. 2019), we set its B/K priors to  
175 medium.

176

177 The intermediate year is a year in the development of a fish stock  
178 such as when biomass, exploitation or recruitment was low or high  
179 (Froese et al. 2019). We set the intermediate years for the stocks  
180 at years when biomass was highest i.e. 1989 for Nile perch, 1991 for  
181 Nile tilapia and 2000 for Dagaa. For the Nile perch and Nile tilapia  
182 stocks, the intermediate years were at the time fishery development  
183 was declining (Figure 1), prompting us to set the B/k priors to  
184 strong depletion (Table 1. For Dagaa, medium depletion was selected.

185

186 In the scenarios of CPUE as the index of abundance, we set the B/k  
187 priors at the start (1999) for the whole lake and Ugandan part at  
188 strong depletion for Nile perch and Nile tilapia, and medium  
189 depletion for Dagaa based on guidance from trends in biomass. The  
190 end year B/K priors were maintained as above for both spatial  
191 scales. The intermediate years for the Ugandan part were set at 2005  
192 for all the stocks, corresponding to a period when fishing effort,  
193 catch, and CPUE were highest or lowest. The corresponding  
194 intermediate B/k priors were set at strong depletion for Nile perch  
195 and Nile tilapia, and medium depletion for Dagaa. This was similar  
196 for the whole lake only that the intermediate years were 2005 for  
197 Nile tilapia, 2008 for Nile perch and 2007 for Dagaa.

198 Finally, a recent period of at least 5 years when catch and  
199 abundance were relatively stable or had similar trends is required  
200 for determining catchability coefficient to relate CPUE to biomass.  
201 The default of the last 5 years was selected for all the stocks  
202 except Dagaa in the whole lake whose catches and abundance had  
203 different trends (Figure 1). We chose a period from 2000-2004 when  
204 biomass and catches for Dagaa in the whole lake were stable.

205

206 The CMSY/BSY were implemented in R using the code for the methods  
207 (Froese et al. 2019). Data used are available online (Musinguzi,  
208 2020). Palomares et al. (2018) established an approach of  
209 classifying fish stocks as collapsed, recruitment impaired,  
210 overfished, or healthy, basing on estimates of  $B/B_{msy}$ . This was used  
211 to define the status of the stocks assessed.

212

### 213 **Results and discussion**

214 For the whole lake, the two indices of abundance used returned  
215 comparable estimates of the fisheries reference points because  
216 values in both scenarios were largely overlapping and falling within  
217 each other's confidence intervals (Table 2; supplementary table 1).  
218 For this reason estimates with CPUE as the index of abundance were  
219 adopted for further consideration. In addition, CPUE is the most  
220 preferred for the methods used (Froese et al. 2017) and its results  
221 were more precautionary for most of the stocks. Tables 2 and 3  
222 present the estimates of the reference points and stock status  
223 determined for the whole lake and Ugandan part respectively. Figures  
224 2, 3 and 4 illustrate the reference points and status for Nile perch  
225 in Uganda as an example, with illustrations of other stocks  
226 available in supplementary material 1.

227

228 In Uganda, the stock status of Nile tilapia and Nile perch was  
229 recruitment impaired, and Dagaa health. Nile tilapia stock was  
230 tending more towards collapsed status while Nile perch and Dagaa  
231 stocks were tending more towards overfished status (Table 4). For  
232 the whole lake, the stock status of Nile tilapia, Nile perch and  
233 Dagaa was collapsed, overfished and health respectively. The Nile  
234 perch and Dagaa stocks were respectively tending more towards  
235 recruitment impaired and overfished status (Table 4).

236

237 Our results confirm widespread overfishing for Nile tilapia and Nile  
238 perch. The poor status corresponds to poor fishing practices and  
239 intensive fishing pressure that have characterized the fisheries of

240 the stocks for a longtime (Nyamweya et al. (2020). Njiru et al.  
241 (2007) assessed the two stocks in Kenya and observed high  
242 recruitment overfishing with 98% of Nile perch and 60% of Nile  
243 tilapia landed immature, high fishing mortality rate and degradation  
244 in life history. The degradation in life history in response to  
245 intensive fishing was found to be lake wide (Njiru et al. 2008).

246

247 The high fishing pressure and its persistence in the lake are  
248 consistent with our estimates of exploitation. In Uganda,  
249 exploitation has been above the reference level since 2006 for Nile  
250 perch (Figure 2 bottom right; Figure 3E) and 2003 for Nile tilapia  
251 (supplementary Figure 1 bottom right; supplementary Figure 2E). At  
252 around the same time, exploitation has since been above reference  
253 level for the stocks in the whole lake. As a result, we observed  
254 degradation in stock size depicted in declining  $B/B_{msy}$  (Figure 2 top  
255 right) and  $B/k$  (Figure 3D). See corresponding supplementary figures  
256 for other stocks.

257

258 Exploitation was higher for Nile tilapia stocks hence its worse  
259 status compared to Nile perch (Tables 2, 3 & 4). We presume that the  
260 Nile perch stock status would be worse than it is were it not for  
261 its high fecundity and pseudo protected areas offshore where fishing  
262 is probably restricted by distance and severity of weather. Nile  
263 perch individuals can obtain absolute fecundity of 16.8 million  
264 depending on size (Ogutu-ohwayo, 1988). This contrasts with the Nile  
265 tilapia whose mean absolute fecundity is 837 (Natugonza et al.  
266 2016). The poor status of the two stocks is collectively illustrated  
267 in the Kobe plots which indicate that the stocks are 97.8% (Nile  
268 perch) and 100% (Nile tilapia) unsustainable in Uganda (Figure 4;  
269 supplementary figure 3) and 99.9% (Nile tilapia) and 99.1% (Nile  
270 perch) in the whole lake, requiring urgent management interventions

271

272 Dagaa, unlike the other stocks had health stock status in Uganda and  
273 the whole lake. The health status is depicted in low exploitation  
274 which has mainly been below the reference level (supplementary  
275 Figure 4 bottom right; supplementary Figure 5E; supplementary Figure  
276 13 bottom right; supplementary Figure 14E). The health stock status  
277 cannot be attributed to good management which has been limited in  
278 the lake (Njiru et al. 2007). Indeed, fishing pressure on the stock  
279 has intensified because the number and panels of seines used to  
280 target it have increased while mesh sizes have declined  
281 (Mangeni-Sande et al. 2019). We attribute the better status of the

282 stock to the high resilience of the species (Froese & Pauly, 2019).  
283 Dagua has ability to double its biomass in <15 months and this is  
284 likely the reason it can bounce back from the high fishing pressure.  
285 Nevertheless, management needs to pay attention because the fishery  
286 is 17.7-30.2% unsustainable (supplementary figures 6 & 15) and  
287 tending more towards overfished status in both the whole lake and  
288 Ugandan part (Table 4).

289

290 Our estimates of MSY provide with managers, the fisheries potential  
291 of the stocks under good management. The MSY estimates for Nile  
292 perch and Nile tilapia were more than the most recent catches in the  
293 Ugandan part and the whole lake (Table 5). In Uganda, rebuilding the  
294 Nile tilapia and Nile perch stocks to MSY level could increase the  
295 catches of the stocks by 9.2% and 29.5% above the most recent  
296 catches respectively. At the whole lake level, the same intervention  
297 could increase the catches of the two stocks by 72.8% and 15.1%  
298 respectively (Table 5). To realize these benefits, managers should,  
299 in management objectives include rebuilding biomass of Nile tilapia  
300 and Nile perch to  $B_{msy}$  levels which were in all cases more than the  
301 current (2015) biomass (Table 2, 3 & 4). Estimates of biomass  
302 available for Nile perch since 2017 when enforcement was  
303 strengthened show that this is possible through eliminating illegal  
304 fishing practices (Hydro-acoustics Regional Working Group, 2019).  
305 The mean biomass of Nile perch in the whole lake in 2019 was 816,694  
306 tonnes, with 705,458 tonnes as the lower limit and 940,922 tonnes as  
307 the upper limit. This was more than the biomass in 2015 and 24.4%  
308 less than  $B_{msy}$  (Table 3). In Uganda, the mean biomass was 422,076  
309 tonnes with lower and upper limits of 366,757 and 485,694 tonnes  
310 respectively which was also more than the 2015 biomass and only  
311 11.3% less than  $B_{msy}$  (Table 2). These values indicate that the  
312 current biomass  $B_{msy}$  gap is closing faster in Uganda compared to the  
313 whole lake. The recovery for the whole lake is probably constrained  
314 by Kenya which unlike Uganda and Tanzania, has not strengthened  
315 enforcement. Kenya should copy.

316

317 Unlike Nile tilapia and Nile perch, the MSY estimates for Dagua were  
318 lower than recent catch. This means that much more is being taken  
319 than is supported by standing biomass, the same process that  
320 gradually brought about the observed poor status of the Nile tilapia  
321 and Nile perch stocks.

322



323 **Conclusion**

324 The major objective of management on Lake Victoria should be  
325 rebuilding biomass of Nile tilapia and Nile perch to a level that  
326 can support catches at MSY. Eliminating illegal fishing practices  
327 has proved to be an effective way to achieve this because of the  
328 observed increase in biomass of Nile perch since 2017 when  
329 enforcement in Uganda and Tanzania was strengthened (Hydro-acoustics  
330 Regional Working Group, 2019). Kenya should do the same while  
331 Tanzania and Uganda should strengthen to close the gap between the  
332 current biomass and  $B_{msy}$ .

333

334 After elimination of illegal fishing practices, the next task of  
335 management should be to ensure that catches remain low until biomass  
336 is  $\geq B_{msy}$  for at least three consecutive years (Froese et al. 2017).

337

338 After rebuilding, catches could be increased to MSY although a  
339 precautionary measure according to FAO is to exploit at the lower  
340 boundary of MSY (Tables 2 & 3) to guard against inefficiencies in  
341 enforcement and natural dynamics of fish stocks (Caddy et al. 1984).  
342 The precautionary measure could also cater for uncertainties in data  
343 due to unreported catches and cross border fishing and fish trade  
344 which are common on Lake Victoria (Heck et al. 2004). Cross border  
345 fishing and trade could lead to uncertainties in estimates of  
346 reference points especially at country level. For example, Kenyan  
347 fishers and traders who confessed to extensive cross-border fishing  
348 and trade give Kenya more catches than expected, a source of  
349 uncertainty (Geheb, 1997; Matsuishi et al. 2006). Indeed, our MSY  
350 estimates for Nile perch and Nile tilapia in Uganda (Table 3) are  
351 trumped by MSY estimates of 86,096 tonnes and 27,892 tonnes for the  
352 stocks respectively in Kenya (Aura et al. 2020). Cross border  
353 fishing and trade is the only plausible explanation for this.

354

355 To facilitate these interventions, routine data collection,  
356 preferably at an annual scale is indispensable to monitor biomass  
357 and catches to support stock assessments such as this to guide and  
358 evaluate management measures. Data collection has been a challenge  
359 particularly in Uganda. For instance, since 2015, no catch  
360 assessment surveys have been done in the Ugandan part of Lake  
361 Victoria which is regrettable.

362

363 **Acknowledgement**

364 We are grateful to funders who supported projects from which the  
365 data used in this study were generated.

366

367 **References**

- 368 Aura, C. M., Nyamweya, C. S., Owili, M., Gichuru, N., Kundu, R.,  
369 Njiru, J. M., & Ntiba, M. J. (2020). Checking the pulse of the  
370 major commercial fisheries of Lake Victoria Kenya, for  
371 sustainable management. *Fisheries Management and Ecology*,  
372 27(4), 314-324.
- 373 Caddy, J.F. (1984) An Alternative to Equilibrium Theory for  
374 Management of Fisheries. In FAO (1984) Papers presented at the  
375 Expert Consultation on the regulation of fishing effort  
376 (fishing mortality). Rome, 17-26 January 1983. A preparatory  
377 meeting for the FAO World Conference on fisheries management  
378 and development. FAO Fisheries Report No. 289 Suppl. 2, 214 p.
- 379 Cooke, S. J., Allison, E. H., Beard, T. D., ... & Welcomme, R. L.  
380 (2016). On the sustainability of inland fisheries: Finding a  
381 future for the forgotten. *Ambio*, 45(7), 753-764.  
382 <https://doi.org/10.1007/s13280-016-0787-4>
- 383 FAO, MSU. (2015). The Rome Declaration: Ten Steps to Responsible  
384 Inland Fisheries. <http://www.fao.org/3/a-i5735e.pdf>. Rome,  
385 Italy.
- 386 FAO. (2020). The State of World Fisheries and Aquaculture 2020.  
387 Sustainability in action. Rome.  
388 <https://doi.org/10.4060/ca9229en>
- 389 Froese, R. and D. Pauly. Editors. 2019. FishBase.  
390 World wide web electronic publication. [www.fishbase.org](http://www.fishbase.org),  
391 version (12/2019).
- 392 Froese, R., Demirel, N., Coro, G. and Winker, H. 2019. A Simple User  
393 Guide for CMSY+ and BSM (CMSY\_2019\_9f.R).  
394 <http://oceanrep.geomar.de/33076/> in December 2019. Accessed  
395 February 2020.
- 396 Froese, R., Demirel, N., Coro, G., Kleisner, K. M., & Winker, H.  
397 (2017). Estimating fisheries reference points from catch and  
398 resilience. *Fish and Fisheries*, 18(3), 506-526.
- 399 Geheb K. (1997) The Regulators and the Regulated: Fisheries  
400 Management, Options and Dynamics in Kenya's Lake Victoria  
401 Fishery. D. Phil. Thesis, Brighton: University of Sussex, 287  
402 pp.
- 403 Heck, S., Ikwaput, J., Kirema-Mukasa, C. T., Lwenya, C., Murakwa, D.  
404 N., Odongkara, K., ... & Sobo, F. (2004). Cross-border Fishing

- 405 and Fish Trade on Lake Victoria. Fisheries Management Series,  
406 Vol 1, 82pp.
- 407 Hilborn, R., & Walters, C. J. (Eds.). (1992). Quantitative fisheries  
408 stock assessment: choice, dynamics and uncertainty. Springer  
409 Science & Business Media.
- 410 Hydro-acoustics Regional Working Group, 2019. A report of the  
411 lake-wide hydro-acoustic survey. LVFO, Jinja, Uganda
- 412 Mangeni-Sande, R., Taabu-Munyaho, A., Ogutu-Ohwayo, R., Nkalubo, W.,  
413 Natugonza, V., Nakiyende, H., ... & Muwanika, V. B. (2019).  
414 Spatial and temporal differences in life history parameters of  
415 *Rastrineobola argentea* (Pellegrin, 1904) in the Lake Victoria  
416 basin in relation to fishing intensity. Fisheries Management  
417 and Ecology, 26(5), 406-412.
- 418 Marshall, B. E., & Mkumbo, O. C. (2012). The fisheries of Lake  
419 Victoria: past, present and future. Nature & Faune, 26(1), 8-  
420 13.
- 421 Matsuishi, T., Muhoozi, L., Mkumbo, O., Budeba, Y., Njiru, M.,  
422 Asila, A., ... & Cowx, I. G. (2006). Are the exploitation  
423 pressures on the Nile perch fisheries resources of Lake  
424 Victoria a cause for concern? Fisheries Management and Ecology,  
425 13(1), 53-71.
- 426 Mudliar, P. 2018. Fire on the Water: Militarization of Fisheries  
427 Management on Lake Victoria. Retrieved from  
428 [https://securefisheries.org/news/militarization-fisheries-lake-](https://securefisheries.org/news/militarization-fisheries-lake-victoria)  
429 [victoria](https://securefisheries.org/news/militarization-fisheries-lake-victoria)
- 430 Musinguzi, Laban (2020): Data used for determining status and  
431 targets for rebuilding the three major fish stocks in Lake  
432 Victoria. figshare. Dataset.  
433 <https://doi.org/10.6084/m9.figshare.13108289.v1>
- 434 Natugonza, V., Ogutu-Ohwayo, R., Efitre, J., ... & Sharon, N.  
435 (2015). The responses of Nile tilapia *Oreochromis niloticus*  
436 (*L. innaeus*, 1758) in Lake Wamala (Uganda) to changing  
437 climatic conditions. Lakes & Reservoirs: Research & Management,  
438 20(2), 101-119.
- 439 Njiru, M., Kazungu, J., Ngugi, C. C., Gichuki, J., & Muhoozi, L.  
440 (2008). An overview of the current status of Lake Victoria  
441 fishery: Opportunities, challenges and management strategies.  
442 Lakes & Reservoirs: Research & Management, 13(1), 1-12.
- 443 Njiru, M., Nzungi, P., Getabu, A., Wakwabi, E., Othina, A., Jembe,  
444 T., & Wekesa, S. (2007). Are fisheries management, measures in  
445 Lake Victoria successful? The case of Nile perch and Nile  
446 tilapia fishery. African Journal of ecology, 45(3), 315.

- 447 NPA, 2019. Mid-term review of the NRM manifesto 2016 – 2021.  
 448 [http://www.npa.go.ug/wp-content/uploads/2020/06/NRM-Manifesto-](http://www.npa.go.ug/wp-content/uploads/2020/06/NRM-Manifesto-MTR-Report-2016-2021.pdf)  
 449 [MTR-Report-2016-2021.pdf](http://www.npa.go.ug/wp-content/uploads/2020/06/NRM-Manifesto-MTR-Report-2016-2021.pdf). Accessed 10<sup>th</sup> September 2020.  
 450 Nunan, F., Cepić, D., Yongo, E., Salehe, M., Mbilingi, B.,  
 451 Odongkara, K., ... & Owili, M. (2018). Compliance, corruption  
 452 and co-management: how corruption fuels illegalities and  
 453 undermines the legitimacy of fisheries co-management.  
 454 International Journal of the Commons, 12(2).  
 455 Nyamweya, C. S., Natugonza, V., Taabu-Munyaho, A., Aura, C. M.,  
 456 Njiru, J. M., Ongore, C., ... & Kayanda, R. (2020). A century  
 457 of drastic change: Human-induced changes of Lake Victoria  
 458 fisheries and ecology. Fisheries Research, 230, 105564.  
 459 Nyamweya, C., Sturludottir, E., Tomasson, T., Fulton, E. A., Taabu-  
 460 Munyaho, A., Njiru, M., & Stefansson, G. (2016). Exploring Lake  
 461 Victoria ecosystem functioning using the Atlantis modeling  
 462 framework. Environmental Modelling & Software, 86, 158-167.  
 463 Ogutu-Ohwayo, R. (1988). Reproductive potential of the Nile perch,  
 464 Lates niloticus L. and the establishment of the species in  
 465 Lakes Kyoga and Victoria (East Africa). Hydrobiologia, 162(3),  
 466 193-200.  
 467 Palomares, M. L., Froese, R., Derrick, B., ... & Pauly, D. (2018). A  
 468 preliminary global assessment of the status of exploited marine  
 469 fish and invertebrate populations.  
 470 Trewavas, E (ed.). 1983. Tilapiine fishes of the genera  
 471 Sarotherodon, Oreochromis and Danakilia. British Museum  
 472 (Natural History). London. 583p.

473 **Tables**

474 Table 1 Relative stock biomass (B/K) prior ranges used for the  
 475 stocks. The ranges depend on the depletion status of stocks:  
 476 very strong depletion (0.01 – 0.2), strong depletion (0.01 –  
 477 0.4), medium depletion (0.2 – 0.6), low depletion (0.4 – 0.8),  
 478 and nearly unexploited (0.75 – 1.0) (Froese et al. 2019)

Spatial scale	stock	B <sub>start</sub> /k	B <sub>int</sub> /k	B <sub>end</sub> /k
Whole lake (biomass)	Nile tilapia	0.4 – 0.8	0.01 – 0.4	0.01 – 0.2
	Nile perch	0.4 – 0.8	0.01 – 0.4	0.01 – 0.4
	Dagaa	0.4 – 0.8	0.2 – 0.6	0.2 – 0.6
Whole lake (CPUE)	Nile tilapia	0.01 – 0.4	0.01 – 0.4	0.01 – 0.2
	Nile perch	0.01 – 0.4	0.01 – 0.4	0.01 – 0.4
	Dagaa	0.2 – 0.6	0.2 – 0.6	0.2 – 0.6
Uganda	Nile tilapia	0.01 – 0.4	0.01 – 0.4	0.01 – 0.2

---

Nile perch	0.01 – 0.4	0.01 – 0.4	0.01 – 0.4
Dagaa	0.2 – 0.6	0.2 – 0.6	0.2 – 0.6

---

479

480 Table 2. Lake wide estimates when abundance is fishery independent CPUE. Estimates are on based BSM with  
 481 approximate 95% confidence limits in parentheses. Estimates for  $F_{msy}$ , MSY and  $B_{msy}$  are long-term averages  
 482 while others are for the last year in the dataset (2015).

Stock	$F_{msy}$ (1/year)	MSY (1000 tonnes/year)	$B_{msy}$ (1000 tonnes)	B (1000 tonnes)	F (1/year)	Exploitation ( $F/F_{msy}$ )
Nile tilapia	0.0514(0.0308-0.0858)	72.8(52 - 102)	446(297-668)	70.1(36.6-170)	0.531(0.219-1.02)	10.3(1.79-39.9)
Nile perch	0.228(0.135-0.384)	246(206-294)	1080(675-1727)	547(365-774)	0.37(0.262-0.555)	1.66(1.09-3.65)
Dagaa	0.474(0.344-0.652)	461(396 - 536)	973(700-1351)	1112(809-1361)	0.484(0.395-0.665)	1.03(0.781-1.46)

483  
 484  
 485  
 486

487 Table 3. Estimates for the Ugandan part of the lake. Estimates are based on BSM with approximate 95%  
 488 confidence limits in parentheses. Estimates for  $F_{msy}$ , MSY and  $B_{msy}$  are long-term averages while others are  
 489 for the last year in the dataset (2015).

Stock	$F_{msy}$ (1/year)	MSY (1000 tonnes/year)	$B_{msy}$ (1000 tonnes)	B (1000 tonnes)	F (1/year)	Exploitation ( $F/F_{msy}$ )
Nile tilapia	0.082(0.0501-0.134)	19.9((15.7-25.4)	121(80.5-183)	30.4(16.3-53.4)	0.556(0.316-1.03)	6.77(2.09-24.4)
Nile perch	0.136(0.0846-0.219)	74.8(58-96.4)	476(316-716)	206(130-302)	0.25(0.171-0.396)	1.86(1.01-4.52)
Dagaa	0.433(0.312-0.6)	95.4(80.6-113)	220(158-306)	242(164-310)	0.446(0.348-0.656)	1.04(0.738-1.62)

490

491 Table 4 Stock status based on the classification of  $B/B_{msy}$  values by  
492 Palomares et al. 2018. In parentheses is the stock status each of  
493 the stocks is tending more towards.

Stock	$B/B_{msy}$	Stock status
<b>Uganda</b>		
Nile tilapia	0.25(0.134-0.44)	Recruitment impaired (collapsed)
Nile perch	0.433(0.274-0.635)	Recruitment impaired (overfished)
Dagaa	1.1(0.745-1.4)	Healthy (overfished)
<b>whole lake</b>		
Nile tilapia	0.157(0.0821-0.381)	Collapsed
Nile perch	0.507(0.338-0.717)	Overfished (recruitment impaired)
Dagaa	1.14(0.832-1.4)	Healthy (overfished)

494



495 Table 5 Estimates of MSY for the stocks in relation to recent  
496 catches (average for last three years (2013-2015)).

Stock	MSY (1000 tonnes/year)	Recent catch (1000 tonnes/year)	Change relative to MSY (1000 tonnes/year)	Change relative to recent catch (%)
Nile tilapia_UG	19.9	18.2	1.7	9.2
Nile perch_UG	74.8	57.8	17.0	29.5
Dagaa_UG	95.4	119.3	-23.9	-20.0
Nile tilapia	72.8	42.1	30.7	72.8
Nile perch	246	213.6	32.4	15.1
Dagaa	461	519.8	-58.8	-11.3

497

498 **Figure captions**

499 Figure 1. Trends in catches and absolute biomass for the three major  
500 commercial fish species on Lake Victoria. Catches are for the whole  
501 lake. Absolute biomass from Nyamweya et al. (2016).

502

503 Figure 2. Trends in key management aspects of the Nile perch fishery  
504 in Lake Victoria, Uganda. The graphs show catches relative to MSY,  
505 with 95% confidence limits in grey (upper left); predicted relative  
506 total biomass ( $B/B_{msy}$ ) with the grey area indicating uncertainty  
507 (upper right); relative exploitation ( $F/F_{msy}$ ) and corresponding 95%  
508 confidence limits in grey (lower left); and stock status in relation  
509 to  $B/B_{msy}$  as a function of  $F/F_{msy}$  for the first (1965), intermediate  
510 (1989) and final (2015) years of assessment. The 50, 80 and 95% are  
511 Confidence levels around the assessment of the final year.

512

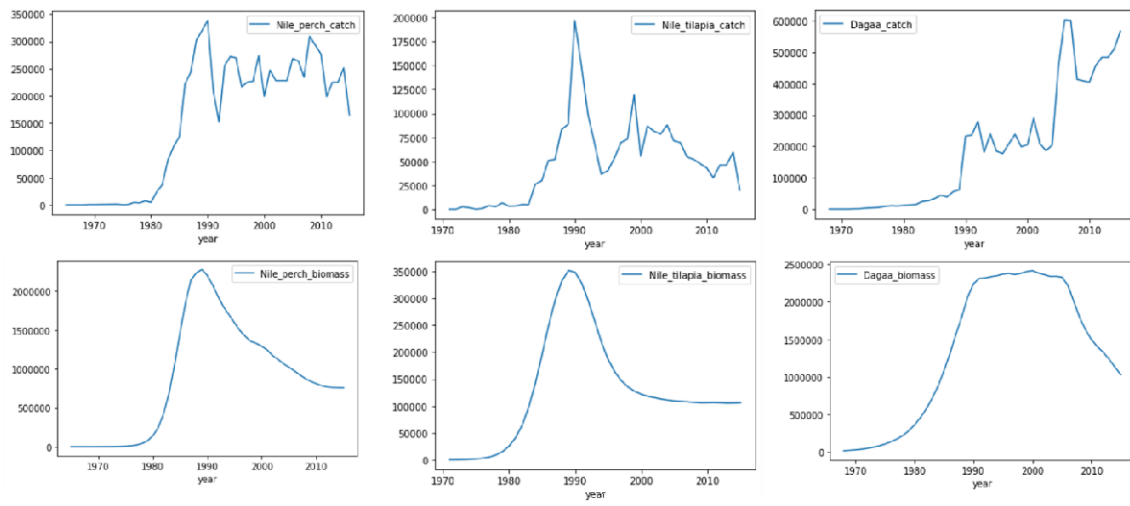
513 Figure 3. Results of the Nile perch fishery in Lake Victoria,  
514 Uganda. A shows time series in catch (black curve) and smoothed data  
515 with indication of highest and lowest catch (red curve). In B to F,  
516 red refers to estimates of BSM and blue to estimates of CMSY+. The  
517 crosses in B show the best r-k estimate of either methods (point in  
518 the center) and their 95% confidence limits (horizontal and vertical  
519 error bars). In dark grey are the pairs found to be compatible with  
520 the catches and biomass. In C, the black and dark grey dots are the  
521 viable r-k pairs found by BSM and CMSY respectively, with indication  
522 of crosses for best estimates with 95% confidence limits. Curves in  
523 D show the BSM and CMSY+ predictions of biomass, the dots the  
524 biomass data scaled by BSM, the vertical blue lines the prior  
525 biomass ranges. E shows the predictions for exploitation and catch  
526 per biomass as scaled by BSM (dots). The curves in F show the BSM  
527 and CMSY+ predictions of Schauer equilibrium curves for catch/MSY  
528 relative to stock size ( $B/k$ ) from the first (square) to the last  
529 year (triangle) of assessment, with the dots showing predicted catch  
530 per predicted biomass as scaled by BSM.

531

532 Figure 4. A Kobe plot for the Nile perch in the Lake Victoria,  
533 Uganda based on CMSY+ estimates of  $B/B_{msy}$  and  $F/F_{msy}$ . A stock in the  
534 orange area is health but vulnerable to depletion by overfishing. In  
535 the red area, a stock is overfished and is undergoing overfishing,  
536 with too low biomass levels to produce maximum sustainable yields  
537 (MSY). In the yellow area, a stock is under reduced fishing pressure  
538 but recovering from too low biomass levels. The green area is the  
539 target area for management, indicating sustainable fishing pressure

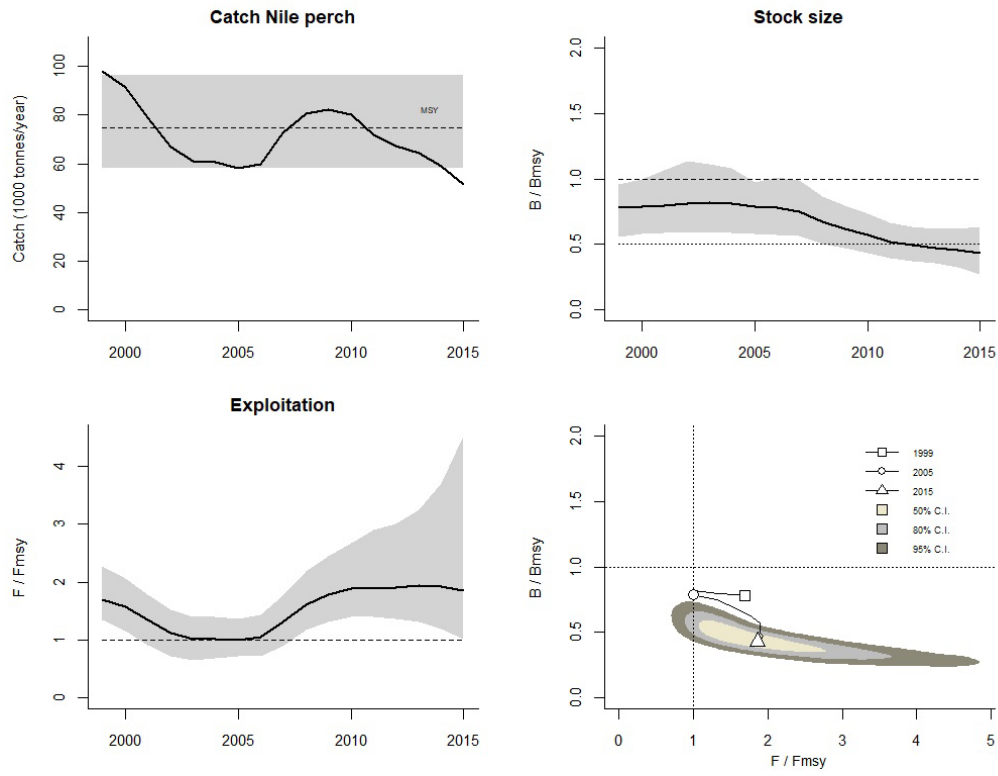
540 and healthy stock size capable of producing high yields close to  
541 MSY. The probabilities of the Nile perch stock being in any of these  
542 areas are given. the last year falling into one of the colored  
543 areas. The 50, 80 and 95% are confidence levels around the year of  
544 final assessment. The legend in the upper right graph also  
545 indicates.

546 **Figures**  
547 **Figure 1**



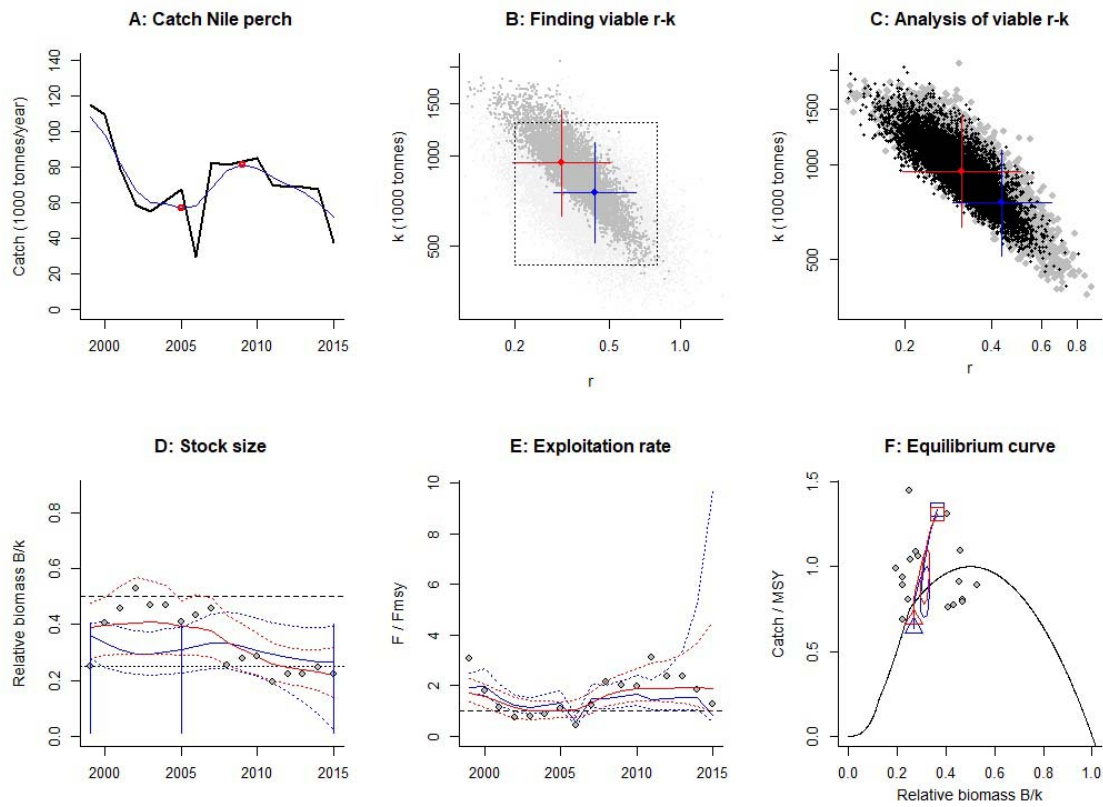
548

549 Figure 2



550

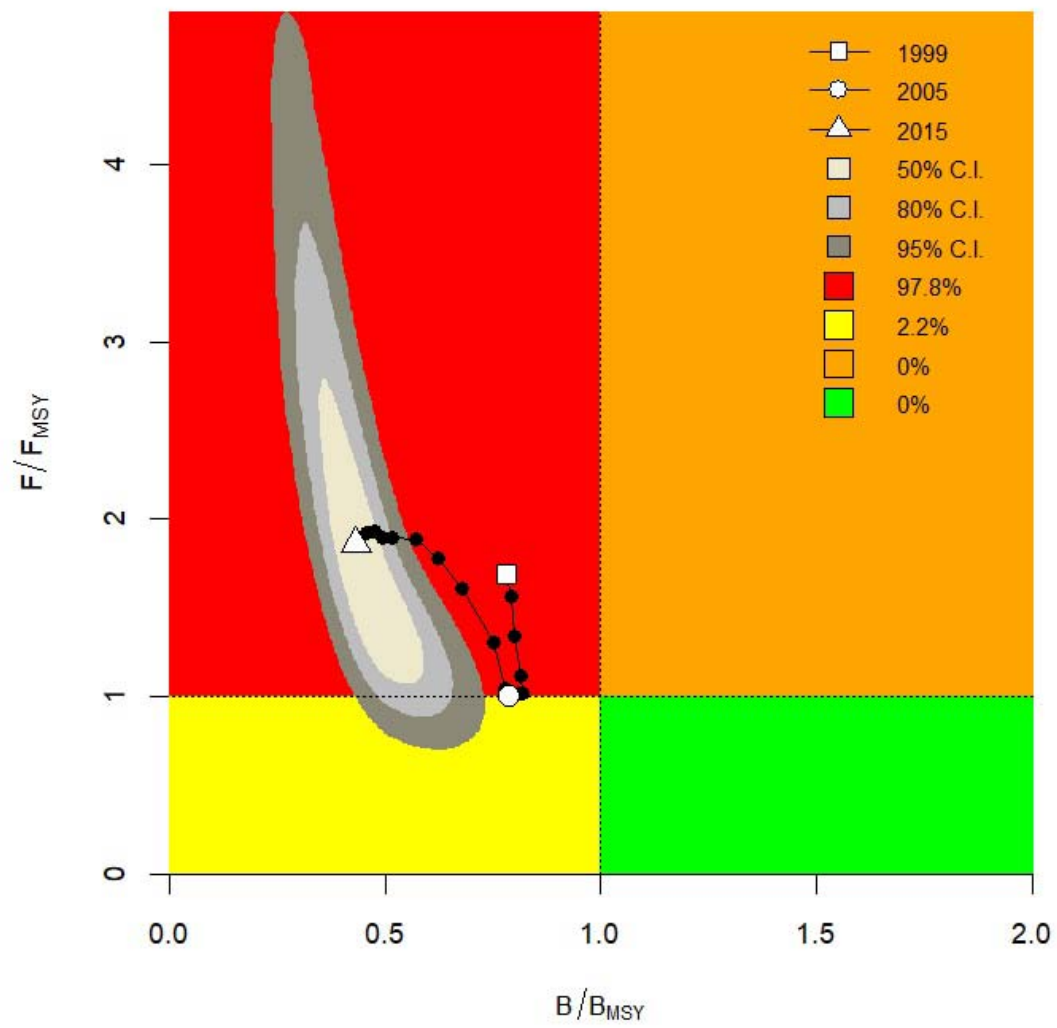
551 Figure 3



552

553 Figure 4

554



555