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Comments on the international consensus model for the water scarcity footprint (AWARE) and proposal for an improvement

Arnaud Helias

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25 area of the region), we show that the DTR-based characterization factors have the same
26 properties than the AWARE-based ones between cut-offs. This article therefore provides
27 a new alternative way of quantifying the impact of water use, in line with the AWARE
28 model features, but without its validity limits and induced thresholds.

29 **Keywords**

30 Life Cycle Impact Assessment; Water Impact; Characterization factor; Marginal
31 approach; Average approach;

32 **1. Introduction**

33 Life Cycle Assessment (LCA) is the generic and global approach dealing with
34 environmental impacts of human activities. Led by an entire community of researchers
35 and practitioners, LCA provides operational assessments of goods and services through
36 a structured framework (Finkbeiner et al., 2006) and guidelines (e.g. European
37 Commission 2013).

38 All comparisons need consensual criteria. Under the umbrella of UN environment, the
39 Life Cycle Initiative leads collective works defining recommended Life Cycle Impact
40 Assessment (LCIA) indicators (Frischknecht and Jolliet, 2016; Jolliet et al., 2018). As part
41 of this, the Available WATER REmaining (AWARE) model has recently been published
42 (Boulay et al., 2018), addressing water issues in LCA.

43 The AWARE model highlights the importance of considering consumption rather than
44 withdrawal and takes into account spatial variability. It results from a massive and
45 collective effort on behalf of the Water Use in LCA (WULCA) working group. The AWARE

46 provides a consensual, operational and recommended indicator for addressing and
47 comparing water impacts, and fully succeeds in this purpose.

48 The present article discusses the shape of the model, as well as associated limitations on
49 its range of validity, which do not distinguish between regions that are more degraded
50 than fair. A subsequent improvement is then proposed. This improvement follows the
51 common practice in LCIA by (1) the definition of a relationship modelling the impact
52 according to human intervention and (2) the use of marginal and average/linear
53 approaches for determining the characterisation factor (CF). This improvement is
54 mathematically sound, all the while satisfying the same expectations as the AWARE
55 model.

56 **2. Methods**

57 2.1. The AWARE model

58 2.1.1. Origin

59 With the purpose of answering the following question, “What is the potential to deprive
60 another freshwater user (human or ecosystem) by consuming freshwater in this
61 region?”(Boulay et al., 2015, 2018), the water impact is logically addressed as first
62 approach using the ratio between the water demand and the water availability (DTA) in
63 a given area.

64 However, Boulay et al. (2015) also highlighted the limitation of the DTA. As the
65 numerator and denominator have the same unit, this ratio is unitless. It does not offer
66 any information concerning the quantity involved, and the DTA obviously cannot be
67 used as a characterisation factor (CF). As an illustration, a 0.1 DTA value could either

68 refer to a demand for 1 m³ over 10 m³ of availability in area A, or to a demand for 1 000
69 m³ over 10 000 m³ in area B. Nevertheless, the use of one same 1 m³ of water in A or in B
70 should not involve the same impact (if using DTA as a CF). The CF must address the
71 “size” of the reserve. In another context, this was precisely the reason for justifying the
72 square of the reserve in the abiotic depletion potential equation (ADP) described in
73 Guinée and Heijungs (1995). This allows for this “size” to be taken into account in the
74 CF.

75 2.1.1. Model

76 The WULCA group investigated several alternatives to overcome the DTA issue. The two
77 most discussed alternatives were DTA_x (which is roughly similar to the ADP without the
78 same rational), which was finally not selected by the group, and the AWARE, which is
79 detailed below.

80 Using a more synthetic notation than that of the initial publication, the AWARE
81 characterisation factor (CF_{AW}) defined in Boulay et al. (2018) is the following:

$$\frac{1}{AMD} = \frac{a}{A - D_E - C_H} \quad (1)$$

$$CF_{AW} = \begin{cases} 0.1 \times \overline{AMD}, & AMD > 10 \overline{AMD} \\ \frac{1}{AMD} \times \overline{AMD}, & 0.01 \overline{AMD} \geq AMD \geq 10 \overline{AMD} \\ 100 \times \overline{AMD}, & AMD < 0.01 \overline{AMD} \end{cases} \quad (2)$$

82 The CF is based on the inverse of the availability-minus-demand (AMD , m³/m².month).
83 The variable a is the area of the region (“area” in Boulay et al. (2018), m²), A the
84 availability (“Availability”, m³/month), D_E the environmental water requirements
85 (“EWR”, m³/month), C_H the human water consumption (“HWC”, m³/month), and \overline{AMD}

86 the global average AMD of freshwater ecoregions where $C_H + D_E < A$ (“ $AMD_{\text{world avg}}$ ”,
87 $\text{m}^3/\text{m}^2\cdot\text{month}$).

88 2.2. AWARE limitations

89 Operationalization of the CF requires consequent work and the WULCA taskforce was
90 committed to collecting all the necessary spatial information in order to determine the
91 variables of the model at a global scale. The human consumption C_H was then used as
92 the (satisfied) human demand. The ecosystem demand D_E was also spatially quantified,
93 but using the environmental water requirement. The definition is therefore different: D_E
94 does not quantify the current consumption by the ecosystem but, rather, the (requested)
95 ecosystem demand, because the ecosystem can only receive what has been left to it. The
96 state of the ecosystem can be categorised under pristine, good, fair or poor conditions
97 (Pastor et al., 2014; Smakhtin et al., 2006). D_E “evaluates minimum water requirements
98 as a fraction of the available flow to maintain freshwater ecosystems in “fair” conditions
99 with respect to pristine flow” (Boulay et al., 2018).

100 A and C_H are estimations of current flows, and $C_H \leq A$. However when a too high human
101 appropriation of water leads to a poor condition ecosystem (i.e. when $C_H > A - D_E$),
102 equation (1) produces a negative result and obviously cannot be used. This is the case
103 with 13% of the global area and up to 33% of world water consumption at a monthly
104 level as indicated in Boulay et al. (2018). In addition, when the ecosystem state reaches a
105 fair condition (when C_H is close to $A - D_E$) equation (1) tends to infinity. Equation (1)
106 therefore needs to be bounded in definition of the CF in equation (2). The WULCA
107 taskforce decided to spread the CF_{AW} over 3 orders of magnitude, between 0.1 and 100
108 times the global average. The upper boundary (when $AMD < 0.01 \overline{AMD}$) excludes 5%
109 of world consumption in addition to the previous 33%. As mentioned by Boulay et al.

110 (2018), the lower boundary ($AMD > 10 \overline{AMD}$) does not have a significant effect (<1% of
111 world consumption) as high AMD result from low C_H .

112 Due to these boundaries, the inverse of the AMD in CF_{AW} is only used for 87% of the
113 world area and 62% of world consumption at a monthly level. The AWARE provides
114 accurate information, expressing the remaining water with respect to the world average
115 for most places. However, it is worth considering a way to overcome this limitation by
116 proposing a relationship that maintains this validity over the whole world and not only
117 for 62% of its water consumption.

118 2.3. New proposal

119 2.3.1. Demand-to-remaining

120 As aforementioned, the present study deals with requested-environmental and fulfilled-
121 human demands. Each should now be addressed separately because of their differences
122 in meaning. First, the initial ideas of the demand-to-availability and availability-minus-
123 demand are considered. The $\frac{D_E}{A-C_H}$ ratio is defined as the demand (requested by the
124 ecosystem) to availability (minus the effective human appropriation), more simply
125 named the demand-to-remaining (DTR). This ratio, considering both ecosystem demand
126 and human consumption, provides useful and straightforward information representing
127 the current state. An arbitrary value of 1 indicates the ecosystem is in a fair condition. A
128 value of 10 implies the ecosystem needs 10 times more water in order to reach a fair
129 condition. Values less than 1 suggest that the conditions are rather good. With this ratio,
130 the state of the ecosystems can be compared, although their surface matters too. For a
131 surface area twice as large, the corresponding impact should be twice more severe, and
132 the DTR ratio therefore has to be multiplied by the area.

133 This leads to the following water impact (I_w , m²), which expresses an area degraded to
134 fair conditions:

$$I_w = a \frac{D_E}{A - C_H} \quad (3)$$

135 2.3.1. Unit change in the model

136 The demands and availability are expressed in m³/month. Flows instead of quantities
137 are consistent with the notion that freshwater is viewed as a flow resource in the
138 classification proposed by Sonderegger et al. (2017). However, the associated
139 elementary flows defined in the life cycle inventory are commonly expressed in terms of
140 quantity (a volume of water, sometimes dated at a given month). This flow (m³) is a part
141 of C_H , although C_H is defined as a flow rate (m³/month).

142 This aspect is not an issue in the AWARE model because the CF is expressed in the world
143 average equivalent and its initial unit no longer appears. As this is not the case when the
144 CF is defined from the DTR, the issue still remains for marginal and average approaches
145 (see below). It can be solved, as equation (3) can be easily defined using demands and
146 availability that are not expressed in terms of flow rate (m³/month), but in terms of
147 quantity (m³) for the given timespan (i.e. m³/month × 1 month). This simply ensures a
148 unit consistency for the definition of CF and does not modify the reasoning and the
149 numerical values used.

150 2.3.2. Characterisation factors

151 Mainly two approaches have been used in LCA to derive CFs, representing a marginal or
152 average change (Hauschild and Huijbregts, 2015). The Life Cycle Initiative guideline
153 recommends using marginal CF when the system under study concerns less than 5% of

154 the issue (Frischknecht and Jolliet, 2016; Verones et al., 2017), while the average CF
155 addresses large changes.

156 The marginal CF ($CF_{DTR,ma}$, m^2/m^3) is the partial derivative of a model of the
157 relationship between the impact and the inventory flow (the marginal change of the
158 impact with respect to a marginal change in the inventoried flow).

$$CF_{DTR,ma} = \frac{\partial I_w}{\partial C_H} = a \frac{D_E}{(A - C_H)^2} \quad (4)$$

159 The average CF ($CF_{DTR,av}$, m^2/m^3) is obtained from the division of the impact by the
160 overall human intervention (Curran, 2017).

$$CF_{DTR,av} = \frac{I_w}{C_H} = \frac{a}{C_H} \left(\frac{D_E}{A - C_H} - \frac{D_E}{A} \right) \quad (5)$$

$$CF_{DTR,av} = a \frac{D_E}{A(A - C_H)} \quad (6)$$

161 For an average CF, the “background” impact has to be removed (corresponding to a state
162 devoid of human intervention $C_H = 0$, see Hauschild and Huijbregts (2015)). This
163 explains the $-\frac{D_E}{A}$ term in equation (5).

164 2.4. Sensitivity of characterisation factors

165 The significance of the approach is addressed by the sensitivity of the CFs according to
166 the components of the model. It can be obtained by the partial derivative to highlight the
167 shape of the relationship and the corresponding equations are available in the
168 supplementary materials. As an illustration, the changes (Δ) in CF values can also be
169 plotted with respect to changes in model parameters. To deal with the behaviour of the
170 AWARE and DTR models, an arbitrary reference set of parameters is therefore defined:

171 $\bar{a} = 1, \bar{A} = 1, \bar{D}_E = 0.45, \bar{C}_H = 0.3$ and $\overline{AMD} = 1$ as well as the corresponding intervals
172 for the first four: $a \in [0.5, 1.5]$ (i.e. $\pm 50\%$), $A \in [0.5, 1.5]$, $D_E \in [0.3, 0.6]$ and $C_H \in$
173 $[0.1, 0.9]$. The interval of D_E represents 30–60% of \bar{A} , which is the limit identified in
174 Boulay et al. (2018). C_H interval boundaries correspond to 10–90% of \bar{A} , indicating the
175 large amplitude in human consumption levels. The parameters are varied one by one
176 within the interval using reference values for the others.

177 3. Results and discussions

178 3.1. Sensitivity of models for comparison purposes

179 Although a non-marginal version has been recently proposed (Boulay et al., 2019), the
180 currently used AWARE CFs are defined as marginal and are thus compared to marginal
181 CFs from DTR. Figure 1 illustrates the changes in the CFs (with respect to the reference
182 point) as a function of the changes of the parameter values (with respect to the
183 reference point for a and A , and as a proportion of A for D_E and C_H).

184 *[Insert Figure 1 about here]*

185 As expected, the CFs of the AWARE and DTR models increase linearly with the area, in
186 the same manner. However, this is only true in the range between boundaries for the
187 AWARE model, whereas this proportional relation remains valid for all values for the
188 DTR.

189 In both situations, an increase in C_H produces the same result as a decrease in A .
190 Although the relationships have an exponential shape, they are actually negative
191 inverse, with order 2 for AWARE and order 4 for DTR (see supplementary materials).
192 The increase grows faster when the AWARE model upper boundary is being reached,

193 and when the complete human appropriation of water ($C_H = A$) is being attained for the
194 DTR model. This implies that the relationships present similar features but at different
195 intervals, without any discontinuities for DTR.

196 The CFs rise along with D_E , displaying a negative inverse relationship for AWARE and a
197 linear relationship for DTR. The trends therefore differ in their shapes but, within the
198 limits identified by Boulay et al. (2018), not particularly in their results. It is noteworthy
199 that, due to the interval boundaries (30–60% of A), the changes in CFs led by D_E are
200 about 10 times smaller than the changes induced by C_H and A .

201 3.2. Characterisation factors according to demands

202 Figure 2 illustrates how the CFs values are determined simultaneously according to
203 demands. This highlights the main contribution of C_H , which drives the value for both
204 models. The ecosystem demand D_E has a lesser effect except for its role in reaching the
205 upper boundary of AWARE CF. The closer $A - C_H$ approaches the threshold D_E , the
206 closer the CF to the cut-off. The DTR model is free from the constraints of this limit, with
207 a continuously increasing CF until complete appropriation of water by human activities.
208 DTR based CFs do not have boundaries, but extreme values are only found with very
209 high C_H (two orders of magnitude between a zero human consumption and 90% of A ,
210 three orders with 99% of A).

211 *[Insert Figure 2 about here]*

212 3.3. Discussions

213 The main properties of CFs based on DTR are listed here, highlighting the similarities
214 and differences with the AWARE. Other alternatives are also briefly mentioned.

215 3.3.1. Meaning of the indicator

216 With the AWARE, a CF that is equal to 10 “directly represents a region where 10 times
217 less water is remaining per unit of surface in comparison to the reference flow, i.e. the
218 world average. It can also be interpreted as a region where 10 times more area time is
219 required to generate the same amount of unused water in comparison to the reference
220 flow”, excluding cut-offs and standardization to the world average, as the “surface-timed
221 equivalent required to generate one cubic meter of unused water in this region” (Boulay
222 et al., 2018).

223 With DTR, a CF equal to $10 \text{ m}^2/\text{m}^3$ corresponds to a 10 m^2 “area” that is degraded to a
224 fair condition by human use of one cubic meter in the region. Although this could be
225 equivalent to 10 m^2 in a fair condition or 1 m^2 in conditions 10 times more severe, it
226 represents the impact expressed in surface equivalent in a fair condition in the region.
227 As the environmental and human demands can be defined in various ways, they are
228 applied differently in the model. Even though the meaning differs, the DTR impact
229 satisfies the same expectation as the AWARE impact, addressing the deprivation for
230 another use due to freshwater consumption in the region.

231 3.3.2. Approach

232 The design process of CFs is different for the AWARE and DTR models. With the AWARE,
233 the CF is directly built considering the desired properties and constraints. One of the
234 constraints is in the dimensionless nature of the DTA ratio, which consequently cannot
235 be used as a CF. This CF is thus defined as marginal. With the DTR, first a model of the
236 impact is established, and properties and limitations are also considered. However, in
237 this case the lack of unit in the ratio is not an issue anymore, precisely because it

238 represents the impact and not the CF. The CFs are consequently defined from the DTR
239 model by marginal and average approaches, which provide CFs with a consistent unit.

240 3.3.3. Model Behaviours

241 The AWARE model addresses in the same way both the human and ecosystem
242 deprivations, by subtracting both C_H and D_E from the availability. The impact covers
243 these two uses of freshwater. The DTR model differs with D_E as numerator and C_H in the
244 denominator. The impact is then mainly focused on the ecosystem issue. However,
245 AWARE and DTR approaches share comparable trends.

246 CFs that are AWARE based within the cut-offs and DTR based show similar behaviours
247 with respect to the area (linear), water availability in the region and human
248 consumption (exponential-like shape, but at a higher order of magnitude for DTR). The
249 relation with respect to the ecosystem demand differs as it is exponential-like and linear
250 with AWARE- and DTR respectively, however the range of applications reduces the gap
251 between them. When there are no cut-offs, AWARE- and DTR-CFs seem quite identical.

252 A non-linearity for D_E could be introduced by considering both the DTRs of the
253 ecosystem and of human demands (see alternatives 1 and 2 in supplementary
254 materials). With these configurations, both the human and ecosystem uses are
255 addressed jointly, as it is done with the AWARE model. Nevertheless, it renders the
256 model more complex without actually modifying its meaning. It therefore seems
257 appropriate to keep the DTR defined here which is simpler.

258 One could also reasonably wish for a linear relationship between CFs and C_H , assuming
259 that a human water appropriation twice larger should induce a two-fold stronger
260 impact. Nonetheless this leads to DTA_x or ADP based CFs (see alternative 3 in

261 supplementary materials). While this kind of relationship is probably better justified by
262 the ADP reasoning (Guinée and Heijungs, 1995), or by considering it as a marginal
263 approach to dynamic stock models (Hélias et al., 2018; Hélias and Heijungs, 2019), than
264 by the argument presented in Boulay et al. (2018), this shape of model was not selected
265 by the WULCA group. By choosing the AWARE, the impact is expected to grow
266 increasingly faster as consumption increases.

267 Looking at CF_{AW} with respect to A or C_H , the upper cut-off induces a “saturation”: above
268 the limit, the CF does not change. This is expected with relationships such as the
269 cumulated normal law or the logistic law (i.e. an exponential limited increase) which has
270 been used in previous models. The WULCA group chose not to introduce this kind of
271 scaling functions due to lack of knowledge on curve tuning parameters. No additional
272 scaling function is thus used in the DTR to overcome the cut-off limitation. It is
273 worthwhile to design the impact (and not the CF) as a logistic function of human
274 consumption, by only using the ecosystem demand for tuning parameters. However, this
275 becomes another topic to discuss.

276 **4. Conclusion**

277 AWARE consensus model brings a major benefit to the community by proposing a
278 shared standard. However, AWARE relationship is only defined when human
279 consumption has spared sufficient water for an ecosystem in fair condition and loses its
280 validity for more severe situations. This leads to the introduction of cut-offs. By defining
281 impact as the fraction of ecosystem demand on what is left by human activity, the DTR
282 model proposed in the present work makes it possible to overcome this limitation.

283 The DTR model is justified by differences in the definition between the demand for the
284 ecosystem (requested quantity to a fair state) and for human use (effective
285 consumption). The marginal and average approaches used on this model lead to
286 corresponding factors. The formal sensitivity analysis showed that the DTR model
287 provides similar features as the AWARE model but can cover all situations. It can
288 therefore be used instead, while guaranteeing the same outcome behaviour without
289 validity limitations.

290 The consensus construction that led to the AWARE model was a necessary and useful
291 task. It resulted with the proposition for a unique midpoint water impact indicator,
292 useful for LCA by practitioners which need consensual and validated impacts. Research
293 in LCA has to take into account two aspects: operationalization by standardization and
294 its improvements in modelling for future use. This work addresses the second aspect in
295 particular.

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298 of this article.

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361

362 **Figure captions**

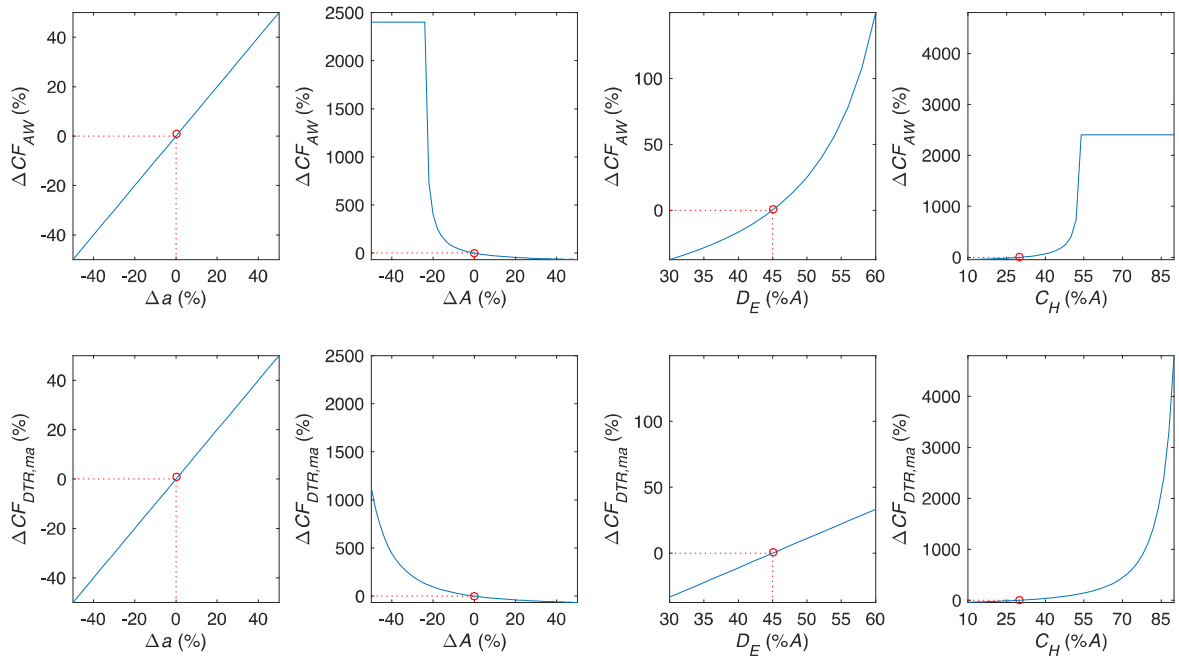
363 Figure 1. Illustration of the sensitivity of the characterization factors according to the
364 model parameters. The circle is the arbitrary reference point.

365 Figure 2. a) AWARE and b) marginal DTR characterization factors as a function of
366 human water consumption ($0 \leq C_H \leq 99\%A$) and ecosystem demand ($60\%A \leq D_E \leq$
367 $60\%A$). Z-axis is in log-scale.

368

369

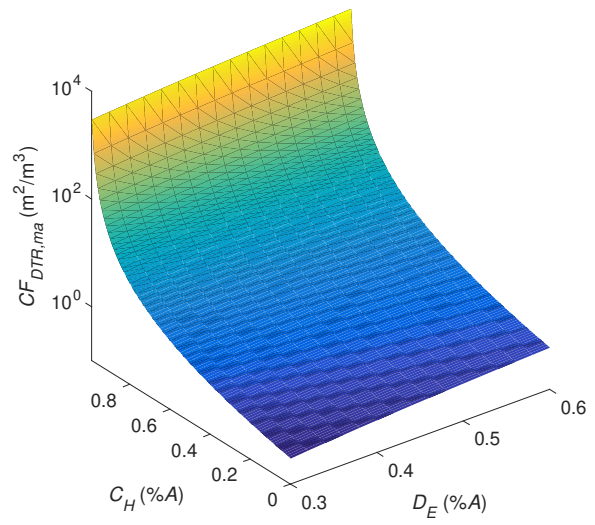
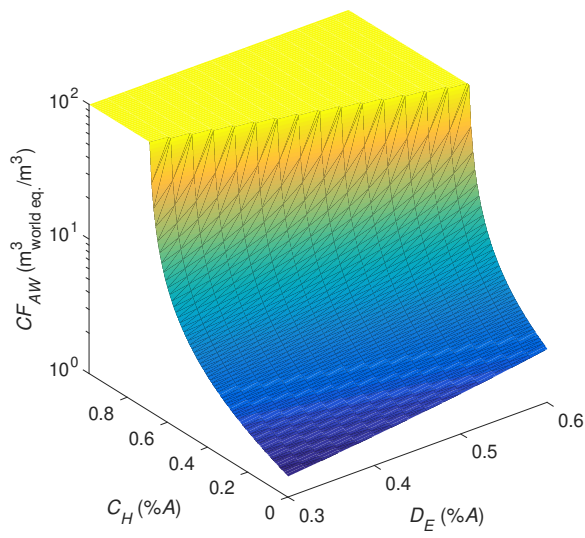
370 Figure 1



371

372

373 Figure 2



374

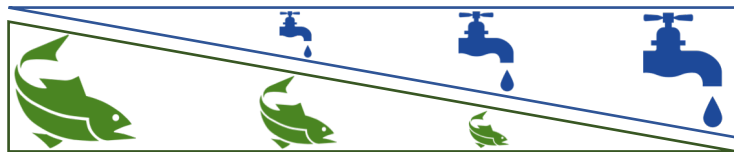
a)

b)

375

376

0% ————— 100% —————> Human water appropriation



Pristine ————— Good ————— Fair ————— Poor —————> State of the ecosystem

