

Replies to editor and associate editor Journal of Hydrology

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Abstract

This document records the replies to the reviewer for the first submission of *Factors determining how catchments respond to forest cover change. Re-analysing global data sets* to Journal of Hydrology, which was rejected before review.

1. Introduction

The reviewer comments are ordered with the Editor in Chief comments listed first and our responses next. This is followed by the comments from the associate editor and our responses. The reviewer comments are in blue and our responses in normal text.

2. Editor comments

The first is to streamline the statistics. As suggested by the AE, a formal model selection process, followed by using only the selected model(s) to evaluate change, would be a suitable approach.

We agree that our statistics was too expansive and not very streamlined. However, we politely disagree with the suggestion of a formal model selection process, as we outline in our reply to the AE. If the statistical modelling was aimed at developing the best predictive model, then this would be the right approach. However, in this case the statistical modelling is aimed at hypothesis testing and explanation of variance in the data set.

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20 In doing so, please carefully note the AEs' concerns about some of the sta-
21 tistical methodology - concerns which range from interpretation to the use of
22 appropriate performance metrics across models of varying structure, to the se-
23 lection of the appropriate metrics of forest cover change for analysis.

24 Please see our comments in reply to the AE below.

25 The second - hopefully supported by a streamlining of the statistics - is to
26 simplify and improve the coherence of the argument. Framing - as the AE states
27 - a "systematic" exploration of the importance of parameters, will improve the
28 readability and interpretability of the work. At present, I think because the
29 statistical analysis itself is convoluted - so too the thread of the argument and
30 clarity of the messages are hard to follow.

31 Please see our comments in reply to the AE below.

32 The final - and perhaps in truth the most problematic issue that may stand
33 in the way of the MS being published - relates to novelty. The AE highlights
34 several recent works with similar questions, approaches and findings at global
35 scales. I could add to that list with more regional studies (e.g. Levy et al 2018).
36 So carefully identifying the knowledge gap being addressed, with respect to
37 these recent studies, and making the case for the present study being "needed"
38 will also be essential.

39 Thank you for the suggestion of the Levy et al. [16] paper, which is an
40 excellent example of a careful statistical analysis taking into account possible
41 variations in climate and dynamic land use change. This is exactly the kind of
42 statistical analysis that we believe needs to be undertaken to better understand
43 how forest cover impacts streamflow. The paper also provides the rainfall and
44 runoff data that was used in the study, but regrettably does not provide the land
45 use data. Rather than deriving this ourselves following the methods described
46 in the paper, we decided that it would be better to use Levy et al. [16] as an

47 example of how the analysis can be done well.

48 You are quite right that we need to be clearer about the objectives of the
49 paper and how this is different from the existing work. We agree that there have
50 been many attempts to derive general conclusions in relation to the impact on
51 streamflow of changes to forest cover, as we also discuss in our introduction.
52 However, as we discuss in our response to the Associate Editor, there are con-
53 siderable issues with the generalization of such studies (as attempted in Zhou
54 et al. [28]; Jackson et al. [14]; Filoso et al. [12] and Zhang et al. [27]). This is
55 now extensively covered in the discussion of the paper.

56 As a result of this, we have rewritten the scope of the paper and changed the
57 title to better reflect the main findings and message coming from this paper.

58 The new title of the paper is:

59 **Generalizing the impact of forest cover on streamflow from exper-**
60 **imental data: it is not that simple.**

61 The key contribution of this paper is to highlight the knowledge gap that
62 exists in the extrapolation of local studies to effects at the global scale. While
63 the impact of forest cover on streamflow is easily hypothesised [e.g. 28, 13] and
64 measured at local scales, our research clearly shows that the global extrapola-
65 tion of the causal relationship between change in forest cover and streamflow is
66 complex and not as straight forward as shown in earlier literature. In addition,
67 to this we highlight that it is very difficult to reinterpret older studies to iso-
68 late the effect of forestation or deforestation, and in many cases this becomes a
69 qualitative assessment.

70 As such, we provide three key insights.

- 71 • While analysing global databases can be interesting, we need to be care-
72 ful with drawing major conclusions (as in Zhang et al. [27], Filoso et al.
73 [12], Zhou et al. [28] and Jackson et al. [14]) based on basic regression

analysis or using equilibrium analysis (such as the Fu model). In many cases statistical assumptions are violated and latent variables can hide or strengthen assumed relationships. In addition, the equilibrium analysis is based on the assumption of water balance closure, which might not always be the case in arid and semi-arid climates. It can easily become a case of ‘correlation without causation’. This is without considering the number of errors that existed in the data. This is particularly important, since results from these global analyses are used to build further models to analyse global impacts [e.g. 13], leading to possible wrong policy or management responses.

- Cumulative and average values of change can be misleading, especially when extracted from published field studies which originally had different objectives. This is particularly true for quite a few studies which focused on regeneration of forests after wildfire or clear cutting followed by re-establishment of plantation of native forest. Many of the Paired Watershed Experiments [e.g. 11, 23, 25, 24, 22] fall in this category, and therefore can easily be classified as either forestation or deforestation. In principle a decision needs to be made how many years post clearing needs to be considered, and whether the remainder of the timeseries should be classified as regeneration. Or, as we discuss, the data should be analysed as a timeseries of change, rather than trying to pick a single point.
- In general, the size of the catchment and the length of the study play a huge role in the interpretation of the results. The length of the study relates to the last point, in all cases, there is large change in the streamflow in the first year, but this effect decreases with the length of the study due to either natural regeneration or some sort of other management, such as replanting. Even though length is not significant in the data, this does

not mean that this effect does not exist in the data. In terms of the catchment size, one of the key issues we originally had with the Zhang et al. [27] paper was the arbitrary split between catchment $> 1000 \text{ km}^2$ and catchment $< 1000 \text{ km}^2$. Our analysis demonstrates that there is no indication of a distinct split, but that, more importantly there is a distinct difference in the type of assessment method reported for the smaller catchment studies (mostly direct observation and paired catchment analysis) and large catchments (mostly hydrological modelling or some sort of statistical modelling). The paper by Beck et al. [4] is an exception, focussing on hydrological model analysis of 12 small catchments in Puerto Rico. However, as we argue, the results of this paper are misrepresented in the database, as none of the analysed catchments had a significant change in the streamflow. This means the change should be recorded as 0.

These are substantial changes and go beyond a major revision. For this reason, we're rejecting the MS at present. If the authors are able to address the 3 issues above in a substantial revision of the MS, we would be pleased to look at it again.

We acknowledge this, and we hope that the current revised version and our responses address these concerns.

We would, however, consider as a new submission for review a substantially revised version of this paper that addresses all of the reviewers' comments. Should you choose to submit such a revised manuscript please refer to the present manuscript number, provide a detailed point-by-point reply to all of the reviewers' comments, and state how the revised manuscript addresses these.

We acknowledge this, and we hope that the current revised version and our responses address these concerns.

127 **3. Associate Editor:**

128 *3.1. Comment AE 1*

129 The manuscript considers an enhanced dataset of streamflow and forest
130 cover, to explore how deforestation/afforestation alters catchment water yields.
131 The manuscript is potentially of interest of the JoH readership, but it is not
132 ready for review in present form.

133 Thank you, we acknowledge this and we hope that our current revision and
134 answers to your comments have improved the manuscript sufficiently to go out
135 for further review.

136 *3.2. Comment AE 2*

137 The main aspects that need to be addressed before the manuscript can be
138 evaluated by experts in the field are listed here. - As apparent from the diag-
139 nostic plots, the model assumptions may be violated in many cases. This can
140 make the results of the fitting (and hence the manuscript conclusions) incorrect.
141 I urge the authors to double check if this is indeed the case and consider ways to
142 address the problem. It is also good practice to check the relevance of outliers
143 (of data with high VIF) and set them aside before model fitting. It is also not
144 correct to comment on models as if working better or worse in certain ranges,
145 based on the residuals (P 29), because the residuals are the results of the data
146 and fitted model, and the fitted model depends on all datapoints.

147 Our answer to this point from the associate editor is quite long and covers
148 the following sub topics:

- 149 • Error distributions
- 150 • The issue of outliers
- 151 • VIF analysis and understanding cross correlations between the variables

152 *3.2.1. Error distributions*

153 Thank you for raising these important points in relation to the validity of
154 the statistical model. A first point that arises from this is that we improve the
155 explanation of the aim of our statistical modelling.

156 There are in essence two approaches to statistical modelling. Generally a model
157 is developed to be used in predictive mode: using a model to predict unknown
158 values, either within or beyond the current data set (forecasting). In this case
159 the model should be reduced to its most efficient version that minimises the bias
160 - variance trade-off. Automatic variable selection and potentially validation on
161 independent data are therefore important, as the aim is to develop the most
162 robust model for prediction.

163 However, a second reason to use a statistical model is to explain the maximum
164 variance in the data. In this case, it is important to develop an a-priori hypoth-
165 esis about the causal relationships in the data. This is subsequently followed by
166 a step by step analysis to test the different causal relationships, either as single
167 variables (as was done in Zhang et al. [27]) or jointly (as in our approach). In
168 this case there is no attempt to find the best predictive model, instead the focus
169 is on the additional amount of explained variance from each variable.

170 This explains why we originally built the model starting from the most simple
171 model, rather than starting from the most complex model.

172 Either way, understanding the diagnostic plots and the residual distribution
173 is important, which is why this was included in the manuscript. In many cases,
174 including such diagnostic plots for single variable regressions is often omitted.
175 For example, both Zhang et al. [27] and Filoso et al. [12] do not present any
176 diagnostics for their relationships and the regressions (Fig 2 - Fig 4 in Zhang
177 et al. [27] and Fig 9 in Filoso et al. [12]) qualitatively indicate issues with the
178 residuals.

179 In our case, we clearly indicate the steps we have taken to improve the quality
180 of the regressions, such as transforming some of the variables, and explaining
181 why we did not take any further steps. Furthermore, the residual distributions
182 are mostly well-behaved, it is only in the tails of the distribution (very high flow
183 changes and very small flow changes) where the residual distribution diverges
184 from normal. As the change in flow variable covers \mathbb{R} , we cannot use a log
185 transformation on the predictant, which is the usual solution for such residual
186 distributions, especially in Hydrology. We therefore chose not to transform and
187 discuss the issue with the fat tails of the distribution in the discussion.

188 We believe that our current discussion of the diagnostic plots in the pa-
189 per clarifies this issue and also highlights the remaining non-normality in the
190 residuals.

191 *3.2.2. The issue of outliers*

192 We agree that outliers could affect the observed residual distribution as this
193 would most likely be obvious in the tails of the distribution. As Venables and
194 Ripley [21] outline on p119: “Outliers are sample values that cause surprise in
195 relation to the majority of the sample. This is not a pejorative term; outliers
196 may be correct, but they should always be checked for transcription errors.” As
197 a result, we believe that excluding values that are outliers is probably not a
198 good idea.

199 However, another careful review of the data identified many problematic
200 values in the data, which were all originally in Zhang et al. [27]. A particular
201 problem was that many catchments had the wrong sign for the change in forest
202 cover. There are many catchments with reported positive change in cover and
203 a large increase in flow. These were all checked and corrected if needed and
204 a full list of all these changes is below and is now included in the paper as
205 Supplementary Data Part 1:

- 206 • 76, Beaver Creek, the flow was corrected from 600% to 157% after review
207 of the original publication [2].
- 208 • 124, D3, Amatya and Skaggs [1]: The originally recorded 250% change by
209 Zhang et al. [27] is clearly wrong. The paper says on page 7: Both of these
210 outflow ratios (0.64 and 0.50) were higher than the calculated expected
211 values of 0.55 for 2003 and 0.44 for 2005, respectively. So the value should
212 be $0.64/0.55 * 100 - 100$ or $0.5/0.44 * 100 - 100$: 16% or 13%. Corrected
213 to 16%
- 214 • 3, Baker Creek, Zhang and Wei [26]. The original recorded 201.1% change
215 by Zhang et al. [27] is also wrong. Original paper says on page 2031:
216 Annual mean flow has been increased by 47.6%. This is now corrected.
- 217 • 67, April rd, which is incorrectly attributed to Ruprecht and Schofield [19]
218 in Zhang et al. [27]. This is actually from Ruprecht and Schofield [18] and
219 the original paper clearly indicates “clearfelling”. As a result the change
220 in forest cover was changed to -100% rather than +100%.
- 221 • 210, March rd, 100, 147.6. Same problem as 67, Bari et al. [3] clearly
222 state that the catchment was cleared, so therefor the change in forest
223 cover changed to -100%.
- 224 • 213, 214 and 215, Monda 1, 2 and 3. These catchments are tricky. The
225 original paper [17] only reports on the control period and indicates that
226 the catchments will be cleared. The later summary paper [22] shows the
227 time series of the flow change, but does not report a single value, so the
228 values in the database must have been estimated from the timeseries. The
229 further complication is that the treatment included clearing and reseedling
230 and regrowth. This suggest that the records should be removed from the
231 database, or only the first few years of the experiment used. In any case, if
232 the values are kept, the sign of the change in forest cover needs to be changed

233 to negative (Clearing).

234 • 230, Oleolega catchment. The paper describes a removal of forest up to

235 85%. changed Delta_F_perc to -85 from 90.

236 • 312, Yerraminup South. The original publication for this catchment is a

237 Western Australian Water Authority report from 1987, which is hard to

238 find, but we have added a copy in the “Papers” folder on github. In this

239 report, in Table 2 on page 11, for the catchment a “Crown cover” decrease

240 of 60% is given. Changed the sign of the change in forest cover: -60%.

241 • 72 Barratta, 100 Coachwood, 103 Corkwood, and 83 Bollygum, as cited by

242 Cornish [10] and Cornish and Vertessy [11]. In the database from Zhang

243 et al. [27], the forest change for all these catchments is positive. However,

244 the paper highlights that these catchments were all logged and either nat-

245 urally regenerated or were planted with a plantation species. So, similar to

246 the the earlier mentioned Monda catchments, the reported change proba-

247 bly only refers to the first couple of years after clearing (before regrowth).

248 In any case, the reported change in forest cover should be negative (clear-

249 ing) rather than positive. Corrected for all three catchments.

250 • 78, Black Spur 1, the treatments and effects are only reported in a con-

251 ference paper [15] and once again indicated clearing, meaning that the

252 change in forest cover should be negative rather than positive (as reported

253 in Zhang et al. [27]). Corrected. Similar to other paired watershed exper-

254 iments, only the first couple of years can be linked to the effect as later

255 regrowth cancels out part of the increase in flow.

256 • 104, Coshocton. Checking the original paper indicates that this is in

257 fact a reduction in flow as a result of reforestation. Changed the sign of

258 Delta_Q_f to be negative.

- 260 • 102, Cold Spring. Checking the original paper [20] indicates that this is
261 in fact a reduction in flow as a result of reforestation. Changed the sign
262 of Delta_Q_f to be negative.
- 263
- 264 • 85 Bosboukloof. This is essentially a duplicate of catchment 184, but
265 the cited paper analyses only 1 year of runoff after a major fire. In any
266 case, the data should reflect a decrease in forest cover: changed the sign
267 of Delta_f_perc to -80%.
- 268 • 259 Shackam Brook. There were a few issues with this catchment in
269 the original database. The name was misspelled and it was incorrectly
270 attributed to Brown et al. [8]. The original paper is the same as 102 [20].
271 Finally, the catchments were all reforestation as the title of the original
272 report indicates and the reported streamflows are all decreases. Corrected
273 Delta_Qf_perc to -20.7%.
- 274 • 95 Sage Brook. Similar to 259 and 102, originates from Schneider and
275 Ayer [20]. Reforestation so Delta_Qf_perc corrected to -19.8%.
- 276 • 101 Coalburn. Original publication (Robinson, 1993) which is a symposium
277 paper, is not available, even after contacting the original authors.
278 The best summary of the research is in Birkinshaw et al. [5] which sum-
279 maries 45 years of research in the Coalburn catchment. It was a refore-
280 station experiment, and there was a decrease in the streamflow over the
281 longer time period. Changed to -20.3%.

282 A further issue was the inclusion of the results of several catchments, for
283 example from the study by Beck et al. [4], which had no significant change in
284 flow. Despite this, the “average” change in flow was reported in the database.
285 We don’t believe that this is correct and the results from such studies should
286 be set to 0. A full list of changes is provided below:

- 287 • 97 Cibucio, 123 Culebrinas, 244 Portugues, 161 Grande de Loiza, 271

288 Tanama, 132 Fajardo, 89 Canovanas, 73, Bauta, 163 Grande de Patillas,

289 283 Valenciano, 181 Inabon, and 162 Grande de Manati. These are all

290 catchments in Puerto Rico from the study from Beck et al. [4]. They

291 should probably be removed from the database as the paper clearly indi-

292 cates that there is no evidence of a change in flow due to reforestation.

293 The values that are cited in the database should all be set to “not signifi-

294 cant from 0”, so might be included as 0. Including them with positive or

295 negative values is misleading. This study is a very detailed hydrological

296 modelling study, but in the end finds no significant change in streamflow

297 as a result of deforestation. Values for all 12 studied catchments set to 0

298 in the database.
- 299 • 188 Kimakia. and 254 Sambret. The data in the database from Zhang

300 et al. [27] appear to originate from Bruijnzeel et al. [9] which gives 3

301 values for different lengths of studies. However, the values in the original

302 study by Blackie [7] and Blackie [6] do not seem to add up to the same

303 values, and the specific values are not mentioned in the actual papers.

304 In addition, as Bruijnzeel et al. [9] mentions in the footnotes, the control

305 for Kimakia is a bamboo catchment, while the control for Sambret is a

306 tea plantation. Overall, this suggests that the data are probably not a

307 clear deforestation/reforestation study and should be discarded from the

308 analysis.
- 309 • 221 N. Creek, Babinda, Queensland. The original paper from this study

310 highlights that the differences between the catchments were insignificant.

311 3.2.3. VIF analysis and understanding cross correlations between the variables

312 We agree that a VIF analysis can be important to identify high correlation

313 between variables. As we have already indicated, we consider Dryness and

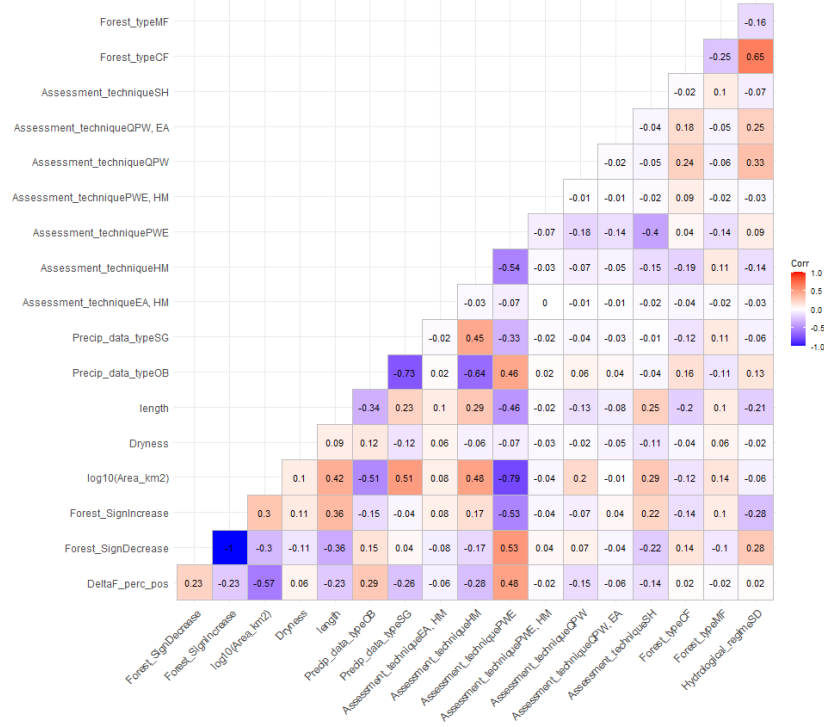


Figure 1: Correlation matrix for all variables

Precipitation to be highly correlated, and therefore we don't include both in the model. However, a VIF analysis is only useful when all the data are numeric, and in this case several of the variables are factors and not numeric.

As an alternative we created a correlation plot (Figure 1). This clearly shows the obvious correlations, but also shows that in general cross correlation is quite low between variables. Some interesting relationships, some of which were also highlighted in our models, appear in this graph:

- the negative relationship between $\log_{10}(\text{Area})$ and change in forest area (DeltaF_perc_pos) indicating that in the data larger catchment tended to have (obviously) smaller areas of forest change.

- the weak positive relationship between $\log_{10}(\text{Area})$ and the assessment method using hydrological models. This is also obvious as it would be impossible to perform paired catchment studies at very large scales.
- As we also indicate in the models, there is a strong inverse relationship between $\log_{10}(\text{Area})$ and the paired watershed assessment method, which is simply the inverse from the last point, as also indicated by the negative relationship between the two assessment methods. This is further visible in the relationship between the change in forest cover and the paired watershed assessment method, showing the impact of the latent variable ($\log_{10}(\text{Area})$). There is of course no causal effect of the assessment methods, it is simply that smaller catchments used in paired watershed assessments are easier to fully clear or fully replant.

Overall this analysis shows very clearly the challenges of simply investigating single variable regressions as was done in Zhang et al. [27] and Filoso et al. [12] or even using simple modelling as in Jackson et al. [14] and Zhou et al. [28]. It is too easy to miss the latent variables that are the underlying factors and influence the model results.

3.3. *Comment AE 3*

- The manuscript presents a number of alternative statistical models, differing by candidate explanatory variables. Each model is designed considering the key shortcomings of the previous one. The end result of such an approach is a complex and somewhat non systematic exploration of predictors and their explanatory power, where it is easy to get lost. I suggest restructuring the manuscript around a well-designed and robustly formalized model selection. One way to proceed could be to start with the most complex model suggested by the extant understanding of the processes at play, and then proceed with a model simplification, according to some consistent criteria (AIC, dropping non

351 significant terms, or similar; high r^2 is not a good criterion because it does not
352 consider the number of parameters). A full blown model selection would also
353 allow to retain or discard the interaction terms, which could be important (as
354 also recognized by the authors; Section 4.5) and should not be discarded a pri-
355 ori. Doing a proper model selection and presenting the results only for the best
356 model (according to a clearly specified criterion) would be less subjective and
357 allow to drastically reduce the number of figures and tables, allowing the reader
358 to focus their attention on the key message.

359 The Associate editor raises several points that need further discussion.

- 360 • A formalised model selection;
- 361 • The use of AIC rather than r^2 for model selection; and
- 362 • Interaction terms.

363 3.3.1. *A formalised model selection*

364 We agree that a classical statistical approach would involve a formal model
365 selection. We acknowledge that in the manuscript we ended up mixing two
366 approaches, where we should have stayed with a single approach. As outlined
367 earlier, the current focus of the statistical modelling is on understanding the
368 different covariates that explain the variation in the data, and to identify latent
369 variables that cause the apparent relationships in the data. This means that
370 there is no attempt to do a formal model selection, as we are not seeking the
371 best predictive model.

372 However, we have written this badly in the original manuscript and have
373 now reworded large sections of the methodology and the results to be more
374 clear about this approach (line 220 - 224 on page 11)

375 *3.3.2. AIC rather than r^2*

376 We actually used the adjusted r^2 in the paper, which does take into account
377 the number of degrees of freedom in the model and therefore can be used to dis-
378 criminate between models, in exactly the same way as the AIC does. However,
379 as the GAM models also provide an AIC, we have for consistency changed our
380 model performance measure to the AIC.

381 *3.3.3. Interaction terms*

382 The issue of interaction terms is a tricky one. As we indicated in the
383 manuscript, we did not include most of the interaction terms as it became a
384 guessing game. While there clearly is cross correlation between the variables
385 and there is potential interaction between terms, the question of clear causal-
386 ity remains unanswered. We therefore believed that including these interaction
387 terms in the model brought us back to the original point we were trying to make:
388 we need to be careful in simply applying models to global data and assuming
389 relationships that might be clouded by latent variables. As a result we have now
390 dropped the interaction term analysis in the paper.

391 *3.4. Comment AE 4*

392 - The novelty of this work needs to emerge more clearly in the introduction.
393 As it looks now, the manuscript could be easily considered somewhat confirma-
394 tory, with respect to most data, approaches and conclusions reached by Zhang
395 et al 2017 and Filoso et al 2017. Furthermore, the introduction needs to be re-
396 arranged, starting with a clear statement of the problem, what we know about
397 that based on previous results, what is missing/how these previous analyses can
398 be improved, and, stemming from these knowledge gaps and/or our understand-
399 ing of the mechanisms, the questions addressed in the work or the hypotheses
400 tested.

401 We agree that this is a valid point and a weakness in the original manuscript.
402 As we also outline in our reply to the Editor, we have rewritten the paper
403 to strengthen the novelty of the work. In particular the paper now focuses
404 more directly on the difficulties in analysing aggregated global data and the
405 importance of latent variables. We point out that while global databases seem
406 to be a great opportunity to understand global trends and interactions, we show
407 that this has significant challenges and can easily lead to questionable results.

408 3.5. *Comment AE 5*

409 There are also some typos and unfinished sentences (e.g., L 142, L 298).
410 Some units are missing (for example those of length of the experiment in the
411 figures) and symbols are not defined at their first appearance (E0/Pa in L 99;
412 Dryness Index). Also: how is Table 1 used? These are not big issues per se but
413 are nonetheless distracting.

414 Thank you for pointing this out, we have reviewed the paper carefully and
415 hopefully corrected these minor issues.

416 3.6. *Comment AE 6*

417 I would also like to provide the authors with a couple of suggestions regard-
418 ing the statistical model and their interpretation. - The models used in the
419 manuscript consider the absolute value of the forest cover change and then its
420 sign, but this choice is not well justified. It implicitly assumes that the status
421 corresponding to no change distinguishes two ‘realms’. Yet, I would expect (and
422 it is also hinted at at some point in the manuscript) that what really matters
423 is the %forested area (possibly in relation to the climatic conditions) and how
424 it changes. So, I would suggest the authors to consider whether a model nearer
425 to our understanding of the phenomena at play would be one including, for the
426 forest part, %change in forested area (with sign) and %forested area, with the
427 latter possibly as random effect, if not of interested.

428 We agree with the AE that this was a logical idea and links back to our point
 429 about “latent variables”. The excellent paper by Levy et al. [16] includes % forest
 430 area as a variable, but their data cannot be incorporated into the current paper.
 431 This is because their analysis focuses on a time progression rather than a “before
 432 and after” or a clear “control” and “treatment”. Regrettably, extracting the total
 433 area of forest from the papers is not a trivial job, and also has some issues. Some
 434 papers actually don’t provide the total forested area, or only a range is given.
 435 Additionally, the actual data is likely to show the same skew as the current
 436 information, with most of the small and paired watershed catchments having a
 437 100% cover and only the larger catchments having mix of landuses.

438 As a test, and to further address the AE’s comments we have collected
 439 as many of the data for the total forest area for the larger catchments (the
 440 catchments $> 1000 \text{ km}^2$ from Zhang et al. [27]). This results in 60 data points
 441 (as not all forest areas were recoverable from the papers). This variable is called
 442 *Perc_Farea_pre* in Table 1. The results of a model that includes this variable
 443 and simply assumes a generalized linear structure, indicates that the percentage
 444 total area of forest in the catchment is insignificant, even though this model
 445 explains 82 % of the variation in the data (for only the larger catchments).

446 Again, it is clear that the assessment technique (in this case there are no
 447 paired watershed experiments) is significant in explaining the variation in the
 448 data. Not surprisingly this is the Hydrological modelling technique.

Table 1: Statistical overview of the linear components of the model
 including percent forested area

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-16.05	15.08	-1.06	0.3
DeltaF_perc	-0.37	0.12	-3.24	0

	Estimate	Std. Error	t value	Pr(> t)
log10(Area_km2)	1.56	2.31	0.67	0.51
Dryness	4.4	3.22	1.37	0.18
Perc_Farea_pre	0.04	0.08	0.49	0.63
Precip_data_typeOB	-19.28	6.55	-2.94	0.01
Precip_data_typeSG	-12	6.76	-1.78	0.09
Assessment_techniqueEA,	1.37	9.86	0.14	0.89
HM				
Assessment_techniqueHM	15.44	6.17	2.5	0.02
Assessment_techniqueQPW	2.8	9.9	0.28	0.78
Assessment_techniqueSH	18.51	7.65	2.42	0.02
Forest_typeCF	-9.47	6.17	-1.54	0.14
Forest_typeMF	0.28	4.64	0.06	0.95
Hydrological_regimeSD	20.98	6.66	3.15	0

449 Inspecting the residuals of the model (Figure 2) also indicates (as we have
450 highlighted earlier) that the data for the larger catchments are much more well-
451 behaved. In our opinion, this is because most of the larger catchment studies
452 are hydrological modelling studies, and what is analysed here is the underlying
453 conceptualised structure of the models rather than real responses of catchments
454 to change in the forest cover.

455 In the manuscript we have not repeated this analysis, but we have included
456 the following text in the discussion (1524 - 1537 on page 30):

457 “One of these latent variables could be the total area of forest in a catchment,
458 as was analysed in Levy et al. [16]. In this case, the total % area of forest was
459 not included in the data. As a test, the total % area of forest for the larger
460 catchments ($> 1000 \text{ km}^2$ in Zhang et al. [27]) were added to the dataset and the

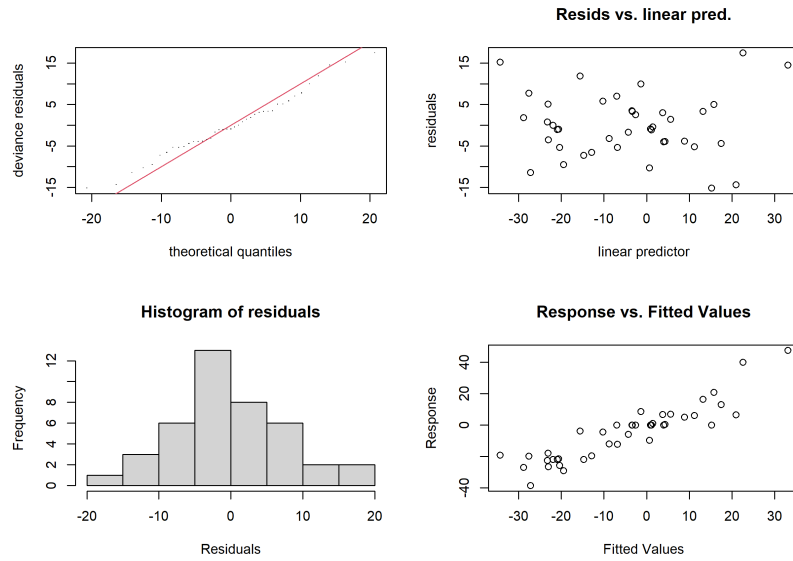


Figure 2: Residual plots for the regression model

model for just the large catchments was tested. This showed that the % area of forest was not significant to explain the change in flow for the larger catchments (retaining all other variables in the model, results not shown). While this might be an area of further research on the full dataset, it is complicated for two reasons:

1. The area of forest is not always indicated in the original papers, or a range of values is given, complicating the data collection.
2. Many of the small catchments have 100% area covered in forest, introducing a strong skew in the data and complicating if total area of forest has an impact on the change in flow.”

3.7. Comment AE 7

- The fact that the explanatory power is low (low r^2) does not necessarily make the results uninteresting (against conclusion on L 530), simply it suggests

474 there are other factors, not included in the model, which have a large effect,
475 and that the model presented cannot be used in a predictive mode. While it
476 is important to present also the r^2 , even a model with low r^2 square we learn
477 which factors significantly affect the change in streamflow and which do not do
478 so.

479 We totally agree with the AE, and this might be more a matter of semantics.
480 We thought we had the same interpretation as the AE, but might have worded
481 this incorrectly. The above comment from the AE actually points exactly to
482 the way we are using models in the paper. We use the models to look at factors
483 that significantly affect streamflow rather than looking at a predictive model.
484 We have reworded the text in the conclusion and have made sure we capture
485 the above suggested meaning.

486 References

- 487 [1] Devendra M Amatya and Wayne R Skaggs. Effects of thinning on hydrol-
488 ogy and water quality of a drained pine forest in coastal north carolina.
489 In *21st Century Watershed Technology: Improving Water Quality and En-
490 vironment Conference Proceedings, 29 March-3 April 2008, Concepcion,
491 Chile*, page 62. American Society of Agricultural and Biological Engineers,
492 2008.
- 493 [2] Malchus B. Baker Jr. Changes in streamflow in an herbicide-treated
494 pinyon-juniper watershed in arizona. *Water Resources Research*, 20
495 (11):1639–1642, 1984. ISSN 0043-1397. doi: [https://doi.org/10.1029/
496 WR020i011p01639](https://doi.org/10.1029/WR020i011p01639). URL [https://agupubs.onlinelibrary.wiley.com/
497 doi/abs/10.1029/WR020i011p01639](https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/WR020i011p01639).
- 498 [3] M. A. Bari, N. Smith, J. K. Ruprecht, and B. W. Boyd. Changes in
499 streamflow components following logging and regeneration in the south-

- ern forest of western australia. *Hydrological Processes*, 10(3):447–461,
1996. ISSN 0885-6087. doi: [https://doi.org/10.1002/\(SICI\)1099-1085\(199603\)10:3<447::AID-HYP431>3.0.CO;2-1](https://doi.org/10.1002/(SICI)1099-1085(199603)10:3<447::AID-HYP431>3.0.CO;2-1). URL <https://onlinelibrary.wiley.com/doi/abs/10.1002/%28SICI%291099-1085%28199603%2910%3A3%3C447%3A%3AAID-HYP431%3E3.0.CO%3B2-1>.
- [4] H. E. Beck, L. A. Bruijnzeel, A. I. J. M. van Dijk, T. R. McVicar, F. N. Scatena, and J. Schellekens. The impact of forest regeneration on stream-flow in 12 mesoscale humid tropical catchments. *Hydrol. Earth Syst. Sci.*, 17(7):2613–2635, 2013. ISSN 1607-7938. doi: 10.5194/hess-17-2613-2013. URL <https://hess.copernicus.org/articles/17/2613/2013/>. HESS.
- [5] Stephen J. Birkinshaw, James C. Bathurst, and Mark Robinson. 45 years of non-stationary hydrology over a forest plantation growth cycle, coalburn catchment, northern england. *Journal of Hydrology*, 519:559–573, 2014. ISSN 0022-1694. doi: <https://doi.org/10.1016/j.jhydrol.2014.07.050>. URL <https://www.sciencedirect.com/science/article/pii/S0022169414005848>.
- [6] JR Blackie. 2.2. 1 the water balance of the kericho catchments. *East African Agricultural and Forestry Journal*, 43(sup1):55–84, 1979.
- [7] JR Blackie. 3.2. 1 the water balance of the kimakia catchments. *East African Agricultural and Forestry Journal*, 43(sup1):155–174, 1979.
- [8] Alice E. Brown, Lu Zhang, Thomas A. McMahon, Andrew W. Western, and Robert A. Vertessy. A review of paired catchment studies for determining changes in water yield resulting from alterations in vegetation. *Journal of Hydrology*, 310(1-4):28–61, 2005. URL <http://www.sciencedirect.com/science/article/B6V6C-4G05MM9-1/2/bbc5fc0e958a8f34bcb7c1cc7fa57b48>.

- [9] Leendert Adriaan Bruijnzeel et al. Hydrology of moist tropical forests and effects of conversion: a state of knowledge review. *Hydrology of moist tropical forests and effects of conversion: a state of knowledge review.*, 1990.
- [10] P. M. Cornish. The effects of logging and forest regeneration on water yields in a moist eucalypt forest in new south wales, australia. *Journal of Hydrology*, 150(2-4):301–322, 1993. URL <http://www.sciencedirect.com/science/article/B6V6C-487D3Y2-9J/2/73c981ba76284d9d629f6b221d6fd6c6>.
- [11] P. M. Cornish and R. A. Vertessy. Forest age-induced changes in evapotranspiration and water yield in a eucalypt forest. *Journal of Hydrology*, 242(1-2):43–63, 2001. URL <http://www.sciencedirect.com/science/article/B6V6C-429910G-3/2/0158b1f89ff436f338a9e688a47f06c4>.
- [12] Solange Filoso, Máira Ometto Bezerra, Katherine C. B. Weiss, and Margaret A. Palmer. Impacts of forest restoration on water yield: A systematic review. *PLOS ONE*, 12(8):e0183210, 2017. doi: 10.1371/journal.pone.0183210. URL <https://doi.org/10.1371/journal.pone.0183210>.
- [13] Anne J. Hoek van Dijke, Martin Herold, Kaniska Mallick, Imme Benedict, Miriam Machwitz, Martin Schlerf, Agnes Pranindita, Jolanda J. E. Theeuwen, Jean-François Bastin, and Adriaan J. Teuling. Shifts in regional water availability due to global tree restoration. *Nature Geoscience*, 15(5):363–368, 2022. ISSN 1752-0908. doi: 10.1038/s41561-022-00935-0. URL <https://doi.org/10.1038/s41561-022-00935-0>.
- [14] Robert B. Jackson, Esteban G. Jobbagy, Roni Avissar, Somnath Baidya Roy, Damian J. Barrett, Charles W. Cook, Kathleen A. Farley, David C. le Maitre, Bruce A. McCarl, and Brian C. Murray. Trading water for carbon with biological carbon sequestration. *Science*, 310(5756):1944–1947, 2005.

- doi: 10.1126/science.1119282. URL <http://www.sciencemag.org/cgi/content/abstract/310/5756/1944>.
- [15] M. D. A. Jayasuriya and P. J. O’Shaughnessy. *The Use of Mathematical Models in Evaluating Forest Treatment Effects on Streamflow*, pages 135–139. Hydrology and Water Resources Symposium, 1988. doi: 10.3316/informit.692214289455295. URL <https://search-informit-org.ezproxy.library.sydney.edu.au/doi/10.3316/informit.692214289455295>. doi: 10.3316/informit.692214289455295.
- [16] M. C. Levy, A. V. Lopes, A. Cohn, L. G. Larsen, and S. E. Thompson. Land use change increases streamflow across the arc of deforestation in brazil. *Geophysical Research Letters*, 45(8):3520–3530, 2018. ISSN 0094-8276. doi: <https://doi.org/10.1002/2017GL076526>. URL <https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1002/2017GL076526>.
- [17] P. J. O’Shaughnessy, K. J. Langford, H. P. Duncan, and R. J. Moran. Catchment experiments in mountain ash forests at north maroondah. *Australian Forestry*, 42(3):150–160, 1979. ISSN 0004-9158. doi: 10.1080/00049158.1979.10674220. URL <https://doi.org/10.1080/00049158.1979.10674220>. doi: 10.1080/00049158.1979.10674220.
- [18] J. K. Ruprecht and N. J. Schofield. Analysis of streamflow generation following deforestation in southwest western australia. *Journal of Hydrology*, 105(1):1–17, 1989. ISSN 0022-1694. doi: [https://doi.org/10.1016/0022-1694\(89\)90093-0](https://doi.org/10.1016/0022-1694(89)90093-0). URL <https://www.sciencedirect.com/science/article/pii/0022169489900930>.
- [19] J. K. Ruprecht and N. J. Schofield. Effects of partial deforestation on hydrology and salinity in high salt storage landscapes. i. extensive

- 578 block clearing. *Journal of Hydrology*, 129(1):19–38, 1991. ISSN 0022-
579 1694. doi: [https://doi.org/10.1016/0022-1694\(91\)90042-G](https://doi.org/10.1016/0022-1694(91)90042-G). URL <https://www.sciencedirect.com/science/article/pii/002216949190042G>.
580
- 581 [20] William Joseph Schneider and Gordon Roundy Ayer. Effect of reforestation
582 on streamflow in central new york. Report 1602, 1961. URL [http://pubs.
583 er.usgs.gov/publication/wsp1602](http://pubs.er.usgs.gov/publication/wsp1602).
- 584 [21] William N Venables and Brian D Ripley. *Modern applied statistics with
585 S-PLUS*. Springer Science & Business Media, 2013.
- 586 [22] Fred Watson, Rob Vertessy, Tom McMahon, Bruce Rhodes, and Ian
587 Watson. Improved methods to assess water yield changes from paired-
588 catchment studies: application to the maroondah catchments. *For-
589 est Ecology and Management*, 143(1):189–204, 2001. ISSN 0378-1127.
590 doi: [https://doi.org/10.1016/S0378-1127\(00\)00517-X](https://doi.org/10.1016/S0378-1127(00)00517-X). URL [https://www.
591 sciencedirect.com/science/article/pii/S037811270000517X](https://www.sciencedirect.com/science/article/pii/S037811270000517X).
- 592 [23] Ashley A. Webb. Streamflow response to pinus plantation harvesting:
593 Canobolas state forest, southeastern australia. *Hydrological Processes*, 23
594 (12):1679–1689, 2009. ISSN 0885-6087. doi: [https://doi.org/10.1002/hyp.
595 7301](https://doi.org/10.1002/hyp.7301). URL [https://onlinelibrary.wiley.com/doi/abs/10.1002/hyp.
596 7301](https://onlinelibrary.wiley.com/doi/abs/10.1002/hyp.7301).
- 597 [24] Ashley A. Webb and Brad W. Jarrett. Hydrological response to wild-
598 fire, integrated logging and dry mixed species eucalypt forest regenera-
599 tion: The yambulla experiment. *Forest Ecology and Management*, 306:
600 107–117, 2013. ISSN 0378-1127. doi: [https://doi.org/10.1016/j.foreco.
601 2013.06.020](https://doi.org/10.1016/j.foreco.2013.06.020). URL [https://www.sciencedirect.com/science/article/
602 pii/S0378112713003885](https://www.sciencedirect.com/science/article/pii/S0378112713003885).

- [25] Ashley A. Webb and Amrit Kathuria. Response of streamflow to afforestation and thinning at red hill, murray darling basin, australia. *Journal of Hydrology*, 412-413:133–140, 2012. ISSN 0022-1694. doi: <https://doi.org/10.1016/j.jhydrol.2011.05.033>. URL <https://www.sciencedirect.com/science/article/pii/S0022169411003519>.
- [26] Mingfang Zhang and Xiaohua Wei. Contrasted hydrological responses to forest harvesting in two large neighbouring watersheds in snow hydrology dominant environment: implications for forest management and future forest hydrology studies. *Hydrological Processes*, 28(26):6183–6195, 2014. ISSN 0885-6087. doi: <https://doi.org/10.1002/hyp.10107>. URL <https://onlinelibrary.wiley.com/doi/abs/10.1002/hyp.10107>.
- [27] Mingfang Zhang, Ning Liu, Richard Harper, Qiang Li, Kuan Liu, Xiaohua Wei, Dingyuan Ning, Yiping Hou, and Shirong Liu. A global review on hydrological responses to forest change across multiple spatial scales: Importance of scale, climate, forest type and hydrological regime. *Journal of Hydrology*, 546:44–59, 2017. ISSN 0022-1694. doi: <https://doi.org/10.1016/j.jhydrol.2016.12.040>. URL <http://www.sciencedirect.com/science/article/pii/S0022169416308307>.
- [28] Guoyi Zhou, Xiaohua Wei, Xiuzhi Chen, Ping Zhou, Xiaodong Liu, Yin Xiao, Ge Sun, David F. Scott, Shuyidan Zhou, Liusheng Han, and Yongxian Su. Global pattern for the effect of climate and land cover on water yield. *Nature Communications*, 6(1):5918, 2015. ISSN 2041-1723. doi: [10.1038/ncomms6918](https://doi.org/10.1038/ncomms6918). URL <https://doi.org/10.1038/ncomms6918>.