Interactive comment on “Impact of humidity biases on light precipitation occurrence: observations versus simulations” by Sophie Bastin et al.

Anonymous Referee #2

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The authors make use of integrated water vapour (IWV) and precipitation from a collection of observations (GPS stations, radio sounding, and combined radar/rain gauge data) in order to evaluate regional climate models operated on various grids (the grid spacing ranges from 0.11° to 0.44°) in a climatological manner (the periods of evaluation cover multiple years to decades). The climate models' simulations are driven by ERA-Interim and participated in the Med-CORDEX initiative. For this purpose, the authors develop a conceptual model that connects IWV, temperature, and precipitation (including Clausius-Clapeyron scaling and deviations from it) that helps to interpret the detected model biases and gives insights in the complexity of precipitation generating processes. From their analyses the authors conclude: 1) all models overestimate
lower values of IWV with an increasing spread among the models during summertime 2) mean biases are mostly explained by model physics (land surface/atmosphere interactions) while dynamics affect the variability 3) the IWV/temperature relationship (that deviates from the Clausius-Clapeyron law) is generally well represented by the models 4) biases in the frequency of occurrence in precipitation can be explained by a higher probability of exceedance of a critical value for IWV (that in turn depends on temperature)

General Comments

There is an endless number of evaluation papers for regional climate models (RCMs) that content themselves with showing biases, but there are only a few papers that deal with the sources of such biases and their underlying processes. The presented manuscript could be one of those rare papers. In addition, the manuscript elaborates on (even if just in a speculative manner) some aspects of the question, how climate change may affect the water cycle. The authors have provided a very interesting and innovative analysis that should be published as soon as possible. However, there are two methodological weaknesses that should to be clarified first, because they may affect the conclusions concerning the interpretation of biases and the precipitation/IWV function (Figure 7) drawn: (1) Comparability between reference (observational) and modelled data The authors make use of RCM data (including re-analysis data ERA-Interim) on various grids and compare it with data from stations (point data) by means of the nearest neighbouring method (cf. page 6, line 12). This has two implications: a) In such a comparison a coarser resolved model is more penalised than a model with a higher resolution, because it has a smaller spatial variability per construction. The coarser resolved model “sees” processes that are resolved by the model with a higher resolution only as “sub-grid scale effects”. As a consequence, a judgement of biases drawn from models on their original grid can be misleading. In order to achieve comparability throughout models of different resolution, modelled and observed data is usually remapped onto a common coarsely resolved grid before the analysis continues
(see Diaconescu et al., 2015; Li and Heap, 2014; Kotlarski et al., 2014). By doing so, it is advisable to recognise the numerical solver of the models: in case of discrete differences, grid cell values are representing averages throughout the grid cell, because of the underlying Reynolds averaging. In such case, a conservative remapping guarantees comparability.  
b) An intrinsic incomparability with IWV data from GPS stations is introduced, because GPS IWV is based on profiles from 4 surrounding ERA-Interim grid cells that are bi-linearly interpolated to the location (latitude and longitude) of the GPS station (Parracho et al., 2018). Hence, the effective resolution of the GPS IWV data is much lower than all models in the manuscript – it is even lower than the IWV from ERA-Interim.

(2) Internal variability and its influence on evaluation results The solution of a local area model is partly predominated by its lateral boundary conditions (LBCs). The larger the model domain or the smaller the grid spacing becomes, the weaker becomes the coupling to its LBCs and the larger become large-scale deviations from its driving data in the interior of the model. Kida et al. (1991) and Paegle et al. (1996) are often cited in this context. More recently, Becker et al. (2015) demonstrated that a local area model creates artificial flows to compensate those deviations in order to achieve physical consistency with the LBCs along the lateral boundaries and that an increase of the model domain does not change this – the artificial flows simply become more complex. As a consequence of this decoupling the model’s variability is increased compared to its driving data. This may lead to added value if the LBCs are derived from a global climate model. However, if the LBCs are taken from ERA-Interim (or some other reanalysis product), the decoupling introduces deviations from observational data. Such deviations are not “wrong”, they just limit the applicability of traditional error statistics. For instance, if there is a thunderstorm at a certain point in time at a certain location in the observations, one cannot expect to find the same thunderstorm at the same location in the model. This has a severe impact on biases that are calculated grid cell by grid cell, but it does not mean, that the model is “wrong” – in a climatological context. This decoupling effect can be seen for instance in Table 3: SD from daily differences
are systematically smaller than SD from 6 hourly data and correlation coefficients on monthly basis are very – although these numbers are affected by issue a). All biases from IPSL20 are systematically larger than those from IPSL50, although both simulations are nudged to large-scale dynamics.

Both methodological issues (comparability and internal variability) have not received any attention yet. However, these issues may severely contribute to the detected biases and their interpretations (which are numerous throughout the manuscript) and hence, they could have significant impact on the conclusions. The authors are kindly asked to revise their analyses, interpretations, and conclusions according to the suggestions below. In order to achieve comparability, all model data should be remapped onto a common grid first and continue with the analysis afterwards. This common grid may depend on the variable, the models, and the reference data. For instance, if IWV from models need to be compared with IWV from GPS, then all models need to be remapped onto the ERA-Interim grid; the final IWV is then derived by a bi-linear interpolation from the 4 surrounding grid cells. In order to avoid misinterpretations of biases stemming from internal variability, one can increase the period and/or the area of averaging. At least, the interpretation of 6 hourly biases should be avoided.


Specific Comments

Page 3, line 38: Parracho et al. (2018) only speaks of 104 GPS stations world wide. How can the authors make use of a hundred of European sites? Page 3, line 39: How accurate are such IWV measurements in the end? Page 5, line 18: ERA-Interim should be referred by Dee et al. (2011). Dee, D. P., Uppala, S. M., Simmons, A. J., Berrisford, P., Poli, P., Kobayashi, S., Andrae, U., Balmaseda, M. A., Balsamo, G., Bauer, P., Bechtold, P., Beljaars, A. C. M., van de Berg, L., Bidlot, J., Bormann, N., Delsol, C., Dragani, R., Fuentes, M., Geer, A. J., Haimberger, L., Healy, S. B., Hersbach, H., Holm, E. V., Isaksen, L., Kallberg, P., Koehler, M., Matricardi, M., McNally, A. P., Monge-Sanz, B. M., Morcrette, J.-J., Park, B.-K., Peubey, C., de Rosnay, P., Tavolato, C., Thepaut, J.-N. and Vitart, F.: The ERA-Interim reanalysis: configuration and performance of the data assimilation system, Q. J. R. Meteorol. Soc., 137(656), 553–597, doi:10.1002/qj.828, 2011. Page 6ff: When models are compared with observational data, the authors simply speak of “differences”. However, it is not always clear how the difference is defined: “model minus observation” (as I suppose) or “observation minus model”? Please, define “difference” somewhere in the methods section and stay with it throughout the manuscript. Page 7, line 39: The explanation of the temperature binning should be explained here. Why is it important that all bins have a similar amount of
data elements? Temperatures do not occur with the same frequency – in fact, it would be easier to follow the argumentation, if the binning would be the same for all models and observations. Page 14, line 29: “The humidity bias thus strongly affects the low precipitation rates, more than the threshold of precipitation triggering.” The latter part of this sentence is inconclusive: 1) a ranking of possible reasons has not been done – however, it would be nice to have. Maybe the authors could explicitly work out this point. 2) Which threshold is meant in this context?

Technical Corrections

The authors are introducing a space character (“ “) prior to a double point (“:”). I find this quite disturbing. It would help, if these unusual space characters could be avoided. Sometimes the LMDZ model is labelled with “LMD”, sometimes it is labelled with “LMD50”. Just for the sake of consistency and also to provide some information about the grid spacing in the acronym, I suggest to use “LMD50” throughout the manuscript. Page 1: line 29: typo: “baises” Page 6, line 23: typo: “Various evaluation metrics . . . has . . .” Page 7, line 39: referring to Figure 7 at this stage is way too early. The numeration of figures and tables should follow the sequence of their first appearance. Page 8, line 10: “This one identifies the minimum value . . .” – shouldn’t it be the maximum? Page 9, line 23: The sentence “This good agreement . . .” is speculative and not relevant for the presented work. Page 9, line 34: typo: “… is a very godd approximation…” Page 10, line 18: “Table 5” – sequence of numeration Page 10, line 18: typo: “… averaged valuers . . .” Page 10, line 29: “ . . . addects . . .” – not clear, what is meant by that; maybe “dominates”? Page 12, line 13: typo “ . . . depiste . . .” Page 12, line 16: typo “ . . . aslo . . .” Page 12, line 33: “ . . . explain important SD . . .” – not clear, what is meant by “important”; maybe “large parts”? Page 13, line 5: “ . . . precipitation picks up . . .” – this phrase is often used in the manuscript. It sounds a bit clumsy. Precipitation more likely “starts to increase”. Page 13, line 11: typo “ . . . the same than . . .” Page 13, line 29: typo “ . . . tendancy . . .” Page 23, Table 3: Are the numbers differences in IWV (as I suppose)? Figure 1, 5c, and 6c: I suggest to
reverse the colours for IWV. When it is more moist (large values) it should be blue. Figure 2: The labels “ReOBS”, “COM 1pc”, and “COM maille” are not defined, here. In more general, it would increase the readability, if the legend and the figure caption would make use of the same acronyms. Figure 2d: Is there a reason for including non-precipitating days in q50? If there are more non-precipitating days than precipitating days q50 simply becomes 0. It would be more informative, if q50 would be based on precipitating days only. Figure 5d: It would be more informative, if Tb1 and Tb2 would be indicated. Figure 6a: What is that red dashed line?