

1. Objective

The impact of climate change on extreme hydrologic characteristics has important socio-economic implications. Regional Climate Models (RCMs) downscale atmospheric information from Global Climate Models to smaller spatial scale, which is then widely used as drivers for hydrological models. Extreme hydrologic phenomena like floods or droughts are often directly linked to extremes in these downscaled meteorological forcing. Consequently, it is important to investigate how reliable RCMs can reproduce extreme statistics of precipitation that have large implications for hydrology.

2. Data Set and Study Area

The RCM data employed in this study was provided by the ENSEMBLES project [3]. The RCMs used for the domain of Germany are shown in Table 1.

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	Name
	C4IRCA3
	CHMIALADIN
	CNRM-RM4.5
	DMI-HIRHAM5
	$ETHZ_CLM$

KNMI-RACMO2 METNOHIRHAM METO-HC HadRM3Q0 **MPI-M-REMO RPN-GEMLAM**

Table 1: Names of RCMs and abbreviation

The RCM runs were conditioned by the ERA40 data set. As reference, the REGNIE product of the German Weather Service (DWD) was aggregated to the spatial resolution of 25 km of the RCMs.

3. Methodology

Principal Component Analysis (PCA) on the correlation matrix of the daily rainfall fields as well as the relative bias for various statistics presented in Table 2 were used to assess the overall performance of all RCMs during the period from 1961 to 2000.

ID	Name	Definition	Unit
Total	Annual Totals	Annual Total precipitation	mm
Rx5	Max 5-day	The maximum of the 5-day	mm
	precipitation amount	precipitation amount	
R10	Number of very heavy	number of days with	days
	precipitation days	more than 10 mm precipitation	
CDD	Consecutive dry days	The largest number of consecutive	days
		days with no more than 1mm rainfall	
R95p	Very wet days	The amount of precipitation higher	mm
		than the 95 percentile of precipitation	
	Table 2: Preci	pitation indices (following [1])	

The Wilcoxon test was used to test the null hypothesis, that the average value of the RCMs (μ_{RCM}) and the observations (μ_{Obs}) for these statistics (Table 2) are equal.

 $H_0 = \{\mu_{RCM} = \mu_{Obs}\}$

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4. Results for the PCA [2]

Number of Eigenvectors

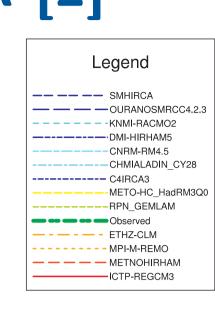
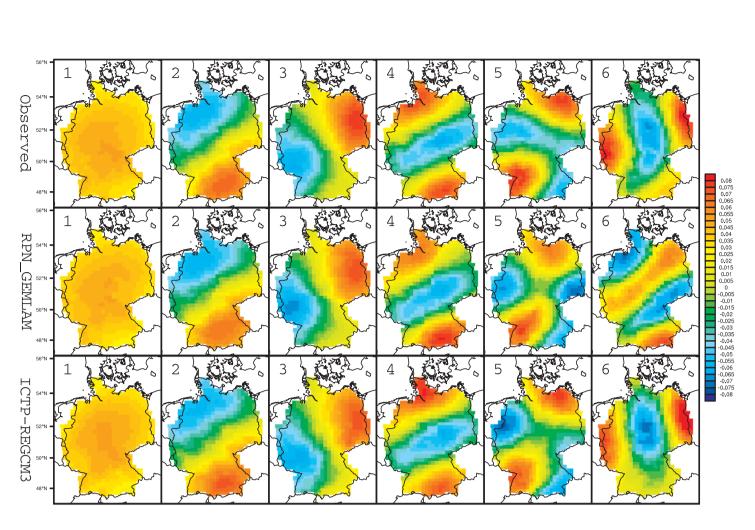


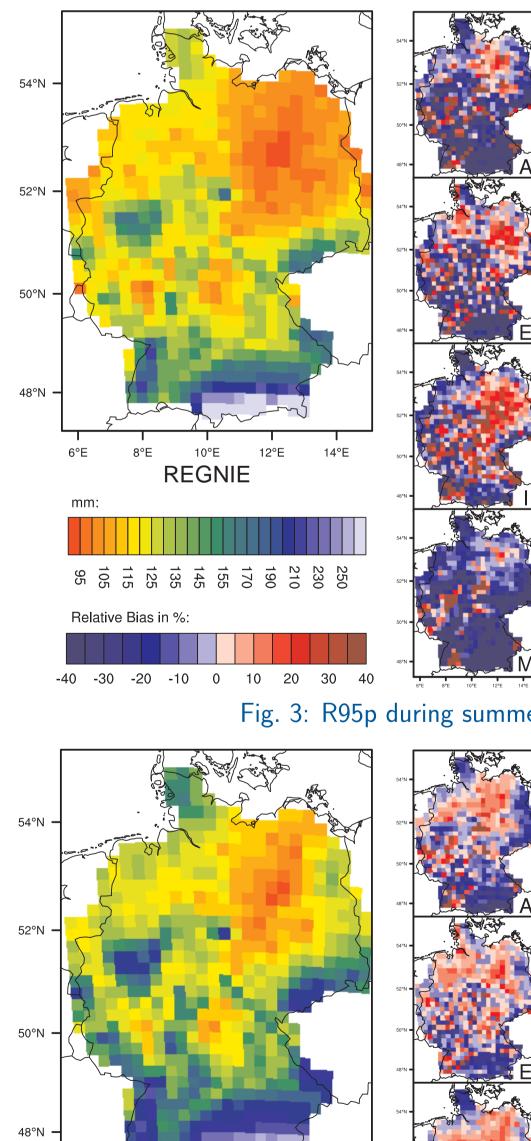
Fig. 1: Cumulative distribution of variance among principal components

The percentage of variance explained by a certain number of principal components (pc) is shown in Fig. 1. In summer, than that obtained in winter. Fig. 2 shows too small.

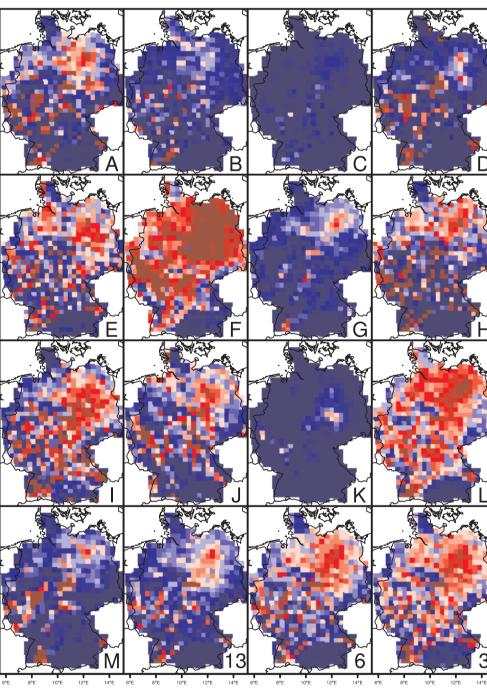


approximately 40 principal components the spatial distribution of the first six are required to explain the 90% of the pcs during summer for the observations, total observed variance, whereas only 20 RPN_GEMLAM, and ICTP-REGCM3. The are required in winter. The spread of the pattern of the sixth pc is better represented cumulative variance explained by the k by ICTP-REGCM3, although its contribulargest eigenvalues is larger in summer tion with respect to the total variance is

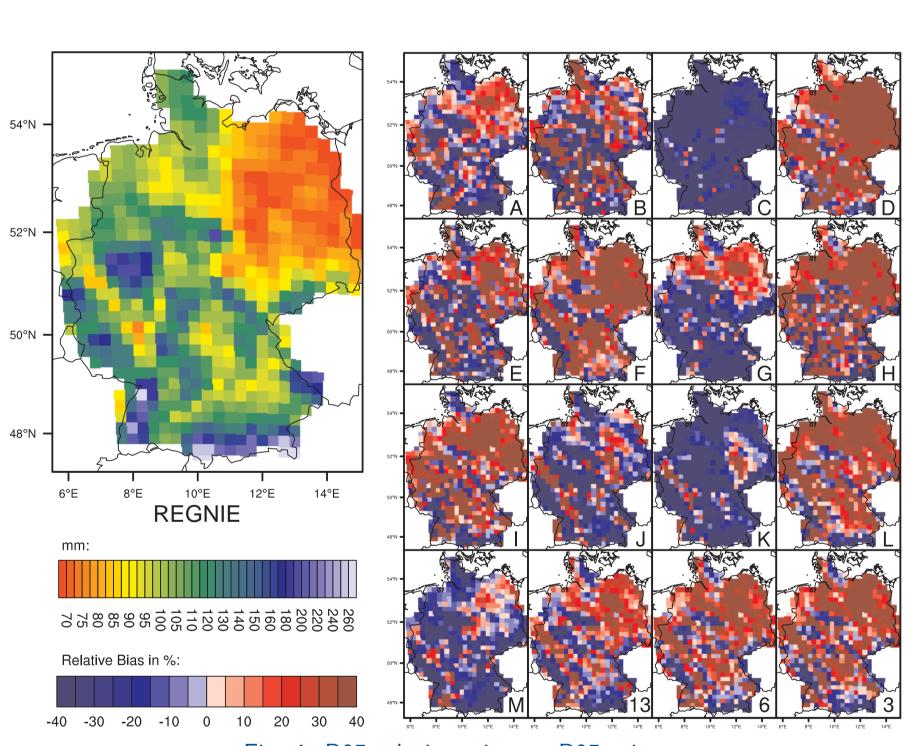
5. Evaluation of Extreme Indices [2]



Relative Bias in %:







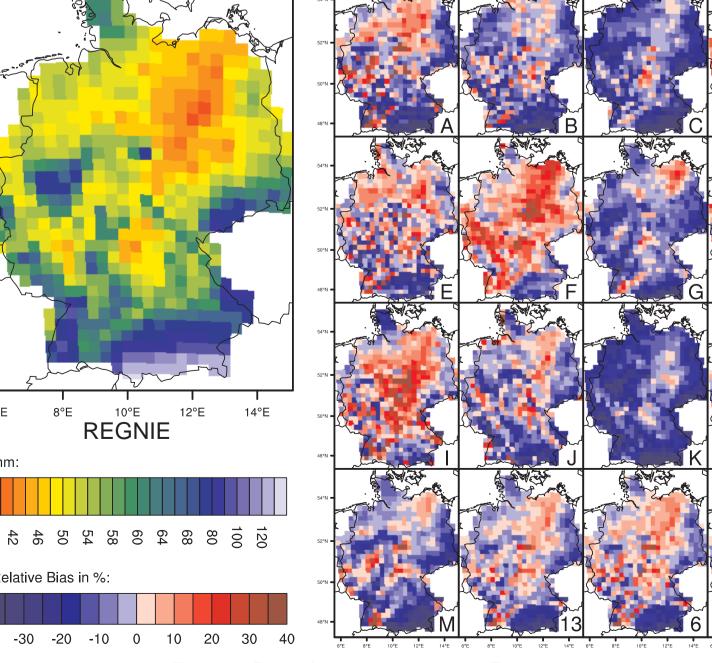


Fig. 5: Rx5 during summer - Rx5sum

Fig. 2: Spatial distribution of eigenvectors

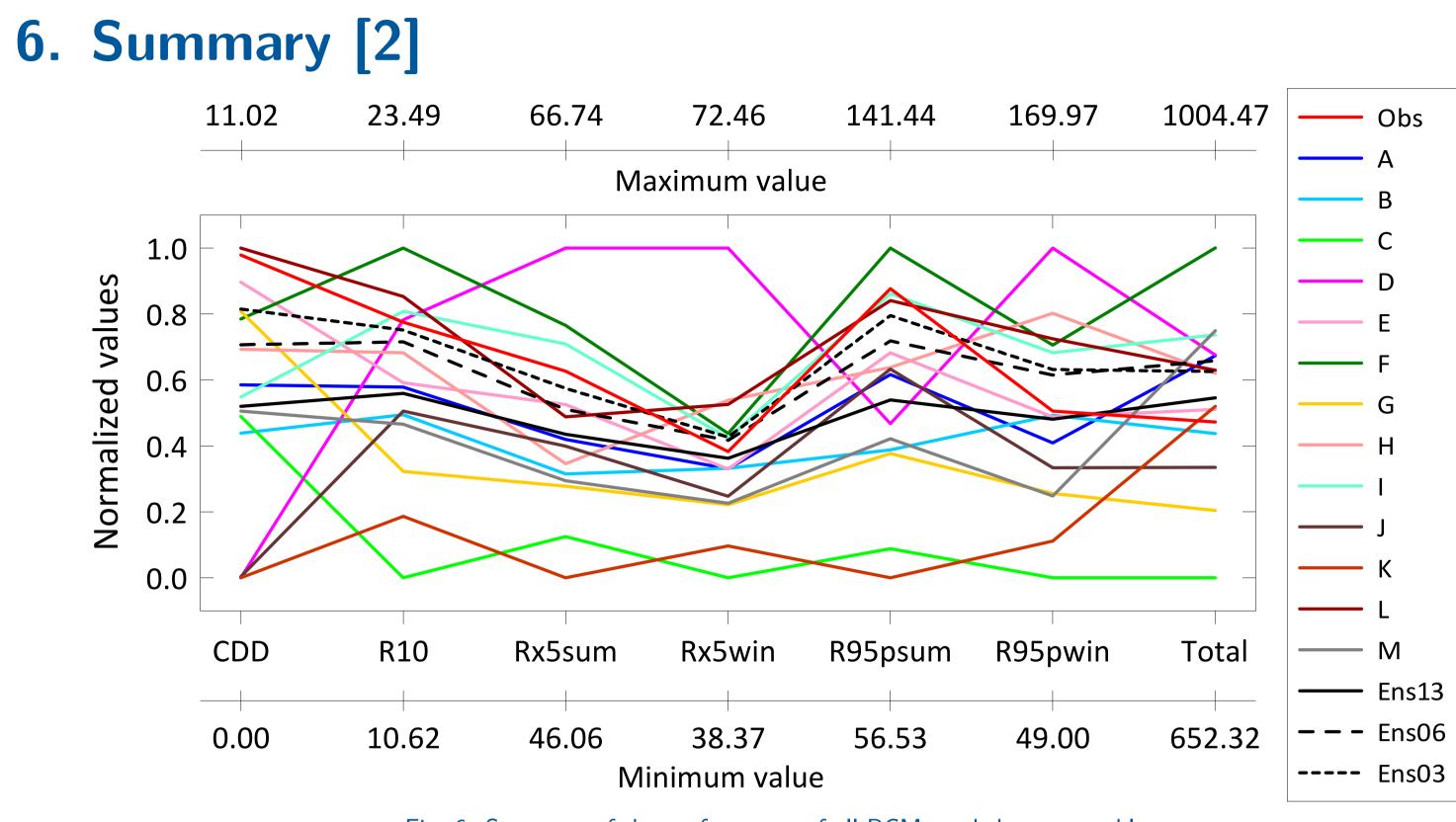


Fig. 6: Summary of the performance of all RCMs and three ensembles

Most RCMs are either under- or overestimating the statistics obtained with the observations. For this reason any 3- and 6- member ensemble performs better than most of the RCMs.

Fig. 4: R95p during winter - R95pwin

The index R95p obtained with summer observations is depicted on the left panel of Fig. 3. The relative bias of each RCM and of the 3-best, 6-best and 13 ensemble members with respect to the latter are shown on the right panel of Fig. 3. Figs. 4 and 5 correspond to the indices R95p in winter and Rx5 in summer, respectively. All models are underestimating R95p and Rx5 during summer in the Alps, but overestimate these indices in the northeast part of Germany with the exception of models C and K. This indicates that the spatial variability is underestimated by most RCMs, since the gradient of both indices is in the northeast - southwest direction.

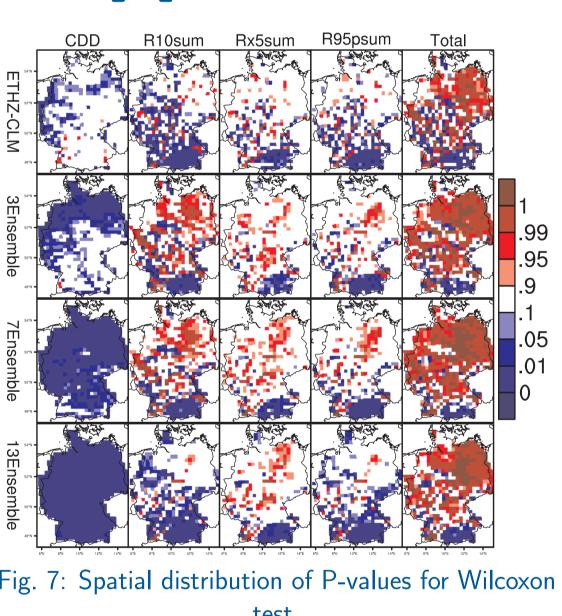
7. Results for the Wilcoxon test [2] The level of rejection of the null hypothesis as well as its spatial distribution is depicted with blue and red colors in Fig. 7, which indicate under- and overestimation, respectively. The number of cells, for which rejection at the 10% significance level was detected, does not decrease for the ensemble statistics and is remarkably high for the annual total precipitation, which is not even an extreme statistic.

8. Conclusions

The PCA indicates that most RCMs are able to preserve the spatial variability of precipitation. Nevertheless, the performance of the RCMs with respect to the selected indices (Table 2) is marked by large biases, which are mostly identified as significant. In general, ensemble with 3, 6 and 13 members outperform all individual models with the exception of ETHZ-CLM.

Consequently, it is not advisable to use coarse RCM data directly for impact assessment studies, because of the highly significant biases detected in all selected extreme indices.

References



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^{2]} S. Thober, L. Samaniego, and R. Kumar, "Evaluation of RCMs: Extreme Precipitation and Temperature

^[3] van der Linden P. and J.F.B.Mitchell, "ENSEMBLES: Climate Change and its Impacts: Su and results from the ENSEMBLES project." p. 160pp., met Office Hadley Centre, FitzRoy